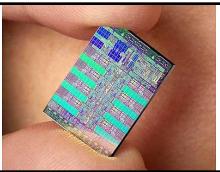
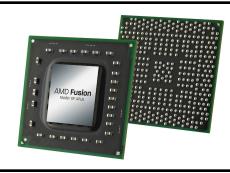
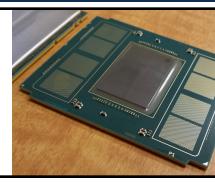
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Profiling Kokkos Applications

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C++ The Bane of Profiling Tools



- It is hard to understand C++ code for a compiler
 - Template Metaprogramming
 - Function Pointers
 - Inheritance
 - Arbitrary aliasing
- It is even harder for a Performance Analysis Tool
 - Most come from a Fortran history
- Abstraction Models make it all worse
 - E.g. only one place where the actual OpenMP loop is
 - Really complex type names which may exceed internal typename length limits
- And we want this across all Platforms ...

Abstractions to the Win?!



- Abstractions can also help us: Instrumentation
- KokkosTools provide built-in instrumentation for Kokkos applications
 - By default enabled on most platforms
- This Instrumentation knows about Kokkos abstractions
 - Get information organized by Kokkos constructs (Parallel Regions, Allocations in Memory Spaces, etc.)
- Enables Meta Instrumentation for Third Party Tools
 - Provide information to Vtune, Nsight, ...
- Easy to use Tools Provide Basic Information accross all Platforms
 - Kernel and Region Times, Memory Utilization, Allocation and Deallocation Frequency, ...

Building an EcoSystem



MiniApps

Applications

Trilinos

(Linear Solvers, Load Balancing, Discretization, Distributed Linear Algebra)

Kokkos – Kernels (Sparse/Dense BLAS, Graph Kernels, Tensor Kernels)

Algorithms (Random, Sort)

Containers (Map, CrsGraph, Mem Pool)

Kokkos (Parallel Execution, Data Allocation, Data Transfer)

std::thread

OpenMP

CUDA

ROCm

(Kokkos aware Profiling and Debugging Tools) Tools Kokkos

Application Support, Developer Training Support Community Kokkos

KokkosTools: github.com/kokkos/kokkos-tools



- Utilities
 - KernelFilter: Enable/Disable Profiling for a selection of Kernels
- Kernel Inspection
 - KernelLogger: Runtime information about entering/leaving Kernels and Regions
 - KernelTimer: Postprocessing information about Kernel and Region Times
- Memory Analysis
 - MemoryHighWater: Maximum Memory Footprint over whole run
 - MemoryUsage: Per Memory Space Utilization Timeline
 - MemoryEvents: Per Memory Space Allocation and Deallocations
- Third Party Connector
 - VTune Connector: Mark Kernels as Frames inside of Vtune
 - VTune Focused Connector: Mark Kernels as Frames + start/stop profiling

How to Use the Tools



- Checkout from github.com/kokkos/kokkos-tools
- Documentation in a Wiki
- Go to src/tools/TOOLNAME and build the tool
 - On most platforms typing "make" is enough
- Before running your code set environment variable
 - export KOKKOS_PROFILE_LIBRARY=/PATH/TO/TOOLS/LIBRARY
- Analyze output
 - Some tools print to screen, some write per-process files
 - Some tools have readers for binary output files
- How does it work internally?
 - Instrumentation is always active, but internal function pointers are NULL
 - At Kokkos::initialize tool library is dynamically loaded, and function pointers are set

Typical Approach



- Run KernelTimer
 - Check if majority of time is in Kernels
 - Check for HotSpot Kernels
- Run MemoryUsage
 - Check where your memory utilization is coming from
 - Check total number of entries to see if frequent alloc/dealloc could be an issue
- If frequent allocations are an issue: run MemoryEvents
 - Figure out which allocations are causing the issue
 - Less than 1000 per second per socket is usually no issue
- To find unaccounted time: put region markers into code
 - Compare region times with kernel times
- Use Connector tools to help investigate individual kernels

The Tools



See Wiki

Exercise



- Use MiniMD a Molecular Dynamics ProxyApp
 - git clone git@github.com:crtrott/miniMD
 - git checkout profiling-exercise
- Basic Molecular Dynamics Importance of Kernels:
 - Force Calculation
 - NeighborList Construction
 - Communication
 - Other stuff: (Integration, Particle Sorting etc.)
- This variant has a problem hidden
 - Use Kokkos Tools to find the issue
 - Follow the typical approach lined out before
 - The main time integration loop is in integrate.cpp::run()
 - Compile:
 - 'make'
 - for CUDA export OMPI_CXX=\${KOKKOS}/bin/nvcc_wrapper
 - Build with fake MPI: got to MPI-Stubs type 'make', build miniMD with HAVE_MPI=no

(NVIDIA) GPU Profiling



- Dominant Performance Bottlenecks:
 - Occupancy
 - Memory Bandwidth
 - Memory Efficiency
 - Memory Load/Store Slots
 - Availability of Instruction Parallelism
- Visual Profiler (nvvp or as part of nsight)
 - Guided Analysis
- Use nvprof to collect data on commandline
 - Generally same information as in the Visual Profiler
 - nvprof [OPTIONS] ./Executable [OPTIONS]
 - --print-gpu-summary : Summary of Kernels, and data transfers
 - --print-gpu-trace: timeline of kernels and data transfers
 - --query-metrics: list of collectable events
 - -m [EVENTS]: set events to be collected
 - --kernels [KERNELS]: restrict profiling to specified kernels

Important Metrics – Occupancy/Mem Sandia National Laboratories

- achieved_occupancy:
 - Actually reached occupancy
 - Cause1: high register pressure (check with --print-gpu-trace)
 - Cause2: high shared memory usage (check with --print-gpu-trace)
 - Cause3: low total available parallelism (too few blocks)
- **_throughput: Bandwidth for different parts of the memory subsystem
 - dram_[read/write]: device memory traffic including ECC
 - [gld/gst]: global memory access, could be cached (this is larger than requested due to efficiency)
 - [gld/gst]_requested: the memory throughput of things the code actually wants
 - I2_I1_read, I2_tex_[read/write],I2_atomic: L2 Cache Throughput by Source
 - local_[load/store]: data traffic due to register spilling
 - shared_[load/store]: shared memory (team scratch level 0)
- **_efficiency: different efficiency metrics
 - [gld/gst]: global memory access (coalesced access = 100%)
 - shared: shared memory loads
- **_hit_rate: Cache hit rates
 - I1_cache_[global/local]: Hit rate in L1 Cache due to global/local load store
 - tex_cache: Hit rate for texture fetches
 - I2_I1_[read]: Hit rate for all L1 misses in L2
 - 12 tex read: Hit rate for texture misses in L2

Important Metrics – Compute



- **_efficiency:
 - sm: multiprocessors are active (small means not enough work launched in a kernel)
 - warp_execution: active threads vs non-active threads due to branching (small means divergence)
 - branch: non-divergent vs total branches (small means divergence in kernel)
 - flop_[sp/dp]: achieved single/double precision peak flop/s fraction
- stall_**: Reasons why a warp does not execute an instruction
 - inst_fetch: instructions are not yet loaded,
 - very unlikely to be an issue, if it is think about breaking up kernel into smaller ones
 - memory_dependency: waiting for a load
 - typical sign for memory bandwidth limitation
 - Use less memory, spread out loads more if possible to overlap with compute
 - exec_dependency: can't execute because prior instruction not done
 - find and expose instruction parallelism
 - memory_throttle: no load/store slots available
 - If this happen without memory_dependency data access patterns are usually bad
 - sync: warps waiting for other warps at a barrier
 - Check if barriers are necessary
 - pipe_busy: compute operation pipe is busy
 - Rarely an issue, except when making haevy use of special function units
 - not_selected: The "good" stall, more ready threads are available than slots are to be filled
 - If you see this as the primary stall reason you have a code which is either artificial or a Gordon Bell candidate

A Case Study I



- A Parametrized Benchmark Code
 - Very fancy vector addition with additional math
- Parameters:
 - N: control total work (RT)
 - M: control data reusage origin (RT)
 - S: control data stride (RT)
 - R: control data reusage (RT)
 - K: control instruction parallelism (CT)
 - F: control Flops/Bytes ratio (RT)

```
View<double***> a("A",N,M,S), b("B",N,M,S),
                 c("C",N,M,S);
// Loop over blocks
for(int i=0; i<N; i++) {
  // Repeat work on block to control cache reusage
  for(int r=0; r<R; r++) {</pre>
    // Loop over block
    for(int j=0; j<M; j++) {</pre>
      a_1 = a(i, j, 0);
      b_1 = b(i, j, 0);
      // Repeat for instruction parallelism
      a_K = a(i,j,0);
      b_K = a(i,j,0);
      // Loop to add more flops
      for(int f=0; f<F; f++) {</pre>
        a 1 += b_1;
        // Repeat for instruction parallelism
        a K += b K;
      c(i,j,0) = a_1 + /*...*/ a_K;
```

A Case Study II



- Exercise I
 - Find settings to measure hardware global and cache bandwidth
- Exercise II
 - Find setting to maximize flop rate
- Exercise III
 - Find settings to make each of the stall reasons the primary stall reasons



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http://www.github.com/kokkos