Introduction to GPU Accelerators and CUDA Programming

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Hands on

- Chunk execution
 - Naive version
 - Stream version
 - Multi-GPU version
- Profiling session





Hands-on Streams: naive version

Write a C or F90 program which performs the following operations:

- Allocate h_A, h_B, h_C single precision arrays of nSize elements on host
- Initializare h_A and h_B arrays using the initArrayData() function in C or the RANDOM_NUMBER() subroutine in F90
- Allocate d_A, d_B, d_C single precision arrays on the *device*
- Transfer data from h_A and h_B arrays on the d_A and d_B arrays
- Launch the arrayFunc() kernel which combine data from d_A and d_B eand write results onto array d_C
- Copy back d_C array from device in h_C array on host
- Measure the total elapsed time to perform both kernel and memory transfers using cudaEvents
- Execute the funcArrayCPU() function which replicates the same CUDA kernel on host for result comparison
- Measure the elapsed time of the funcArrayCPU() function
- Compute the Speed Up of GPU implementation as CPU time / GPU time



Hands-on Streams: naive version





Write a C or F90 program which performs the following operations:

- Allocate h_A, h_B, h_C single precision arrays of nSize elements on host
- Initializare h_A and h_B arrays using the initArrayData() function in C or the RANDOM_NUMBER() subroutine in F90
- Split the elaboration of h_A, h_B arrays into chunks of chunk_size size elements
- Create streams_number of cudaStream
- Allocate d_A, d_B, d_C of chunk_size * streams_number size on the device
- Assign to each cudaStream the elaboration of each chunk. Each stream will:
 - $\hfill \mbox{ copy a chunk of data from h_A and h_B on d_A and d_B buffers }$
 - Launch the kernel arrayFunc
 - Copy back to host the results from d_C into h_C
- Measure execution time and compare the speedup with respect naïve implementation
 - Try to change the number of active streams, the chunk size, etc...











Execution Time Lines

Sequential Version



Asynchronous Versions







CUDA Runtime functions to implement the code (C for CUDA):

- cudaError t cudaStreamCreate(cudaStream t *stream)
- cudaError t cudaStreamDestroy(cudaStream t stream)
- cudaError_t cudaDeviceSynchronize(void)
- cudaErrot_t cudaMemcpyAsync(void* dst, void* src, size_t nbyte, enum cudaMemcpyKind kind,cudaStream t stream)

CUDA Runtime functions to implement the code (CUDA FORTRAN):

integer function cudaStreamCreate(stream)

integer :: stream

integer function cudaStreamDestroy(stream)

integer :: stream

- integer function cudaDeviceSynchronize()
- integer function cudaMemcpyAsync(dst, src, nelements, kind, stream)



Hands-on Streams: cudaStreams and Multi-GPU

Write a C or F90 program which performs the following operations:

- Allocate h_A, h_B, h_C single precision arrays of nSize elements on host
- Initializare h_A and h_B arrays using the initArrayData() function in C or the RANDOM_NUMBER() subroutine in F90
- Split the elaboration of h_A, h_B arrays into chunks of chunk_size size elements
- Assign to each available GPU device a balanced number of chunks to process
- Create streams_number of cudaStream
- Allocate d_A, d_B, d_C of chunk_size * streams_number size on the device
- Assign to each cudaStream the elaboration of each chunk. Each stream will:
 - $\hfill \mbox{ copy a chunk of data from h_A and h_B on d_A and d_B buffers$
 - Launch the kernel arrayFunc
 - Copy back to host the results from d_C into h_C
- Measure execution time and compare the speedup with respect single GPU implementation

