



GPU Teaching Kit
Accelerated Computing



Module 14 – Efficient Host-Device Data Transfer

Lecture 14.3 - Overlapping Data Transfer with Computation

Objective

- To learn how to overlap data transfer with computation
 - Asynchronous data transfer in CUDA
 - Practical limitations of CUDA streams

Simple Multi-Stream Host Code

```
cudaStream_t stream0, stream1;
cudaStreamCreate(&stream0);
cudaStreamCreate(&stream1);

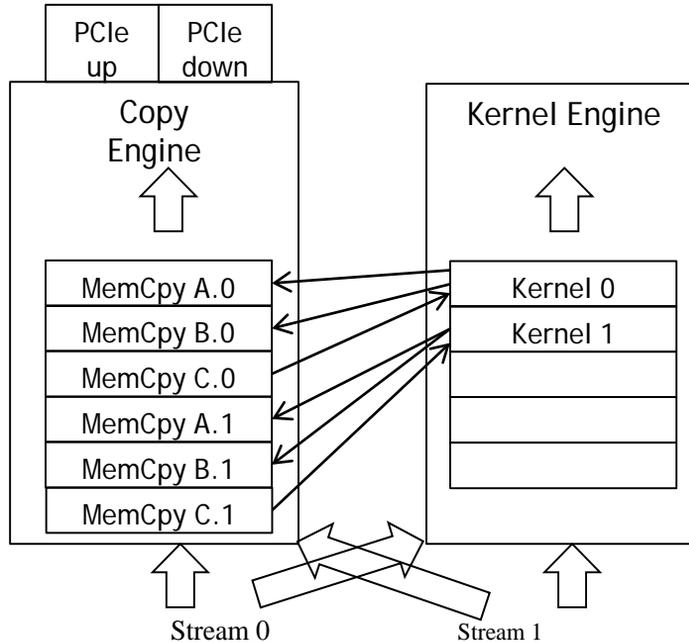
float *d_A0, *d_B0, *d_C0; // device memory for stream 0
float *d_A1, *d_B1, *d_C1; // device memory for stream 1

// cudaMalloc() calls for d_A0, d_B0, d_C0, d_A1, d_B1, d_C1 go
here
```

Simple Multi-Stream Host Code (Cont.)

```
for (int i=0; i<n; i+=SegSize*2) {
    cudaMemcpyAsync(d_A0, h_A+i, SegSize*sizeof(float),..., stream0);
    cudaMemcpyAsync(d_B0, h_B+i, SegSize*sizeof(float),..., stream0);
    vecAdd<<<SegSize/256, 256, 0, stream0>>>(d_A0, d_B0,...);
    cudaMemcpyAsync(h_C+i, d_C0, SegSize*sizeof(float),..., stream0);
    cudaMemcpyAsync(d_A1, h_A+i+SegSize, SegSize*sizeof(float),..., stream1);
    cudaMemcpyAsync(d_B1, h_B+i+SegSize, SegSize*sizeof(float),..., stream1);
    vecAdd<<<SegSize/256, 256, 0, stream1>>>(d_A1, d_B1, ...);
    cudaMemcpyAsync(d_C1, h_C+i+SegSize, SegSize*sizeof(float),..., stream1);
}
```

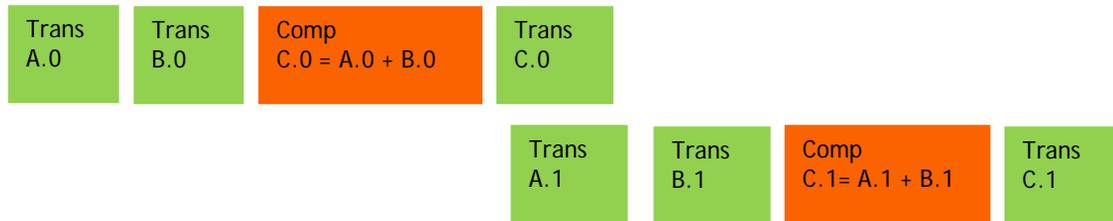
A View Closer to Reality in Previous GPUs



Operations (Kernel launches, `cudaMemcpy()` calls)

Not quite the overlap we want in some GPUs

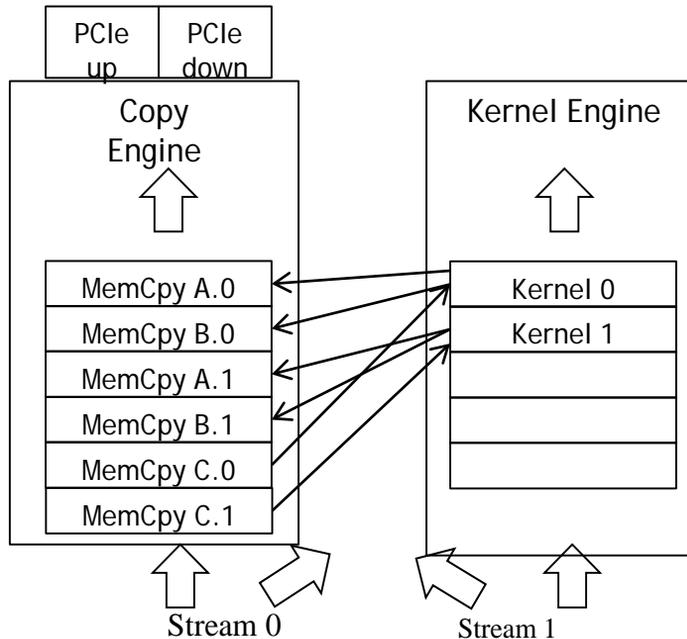
- C.0 blocks A.1 and B.1 in the copy engine queue



Better Multi-Stream Host Code

```
for (int i=0; i<n; i+=SegSize*2) {  
    cudaMemcpyAsync(d_A0, h_A+i, SegSize*sizeof(float),..., stream0);  
    cudaMemcpyAsync(d_B0, h_B+i, SegSize*sizeof(float),..., stream0);  
    cudaMemcpyAsync(d_A1, h_A+i+SegSize, SegSize*sizeof(float),..., stream1);  
    cudaMemcpyAsync(d_B1, h_B+i+SegSize, SegSize*sizeof(float),..., stream1);  
  
    vecAdd<<<SegSize/256, 256, 0, stream0>>>(d_A0, d_B0, ...);  
    vecAdd<<<SegSize/256, 256, 0, stream1>>>(d_A1, d_B1, ...);  
  
    cudaMemcpyAsync(h_C+i, d_C0, SegSize*sizeof(float),..., stream0);  
    cudaMemcpyAsync(h_C+i+SegSize, d_C1, SegSize*sizeof(float),..., stream1);  
}
```

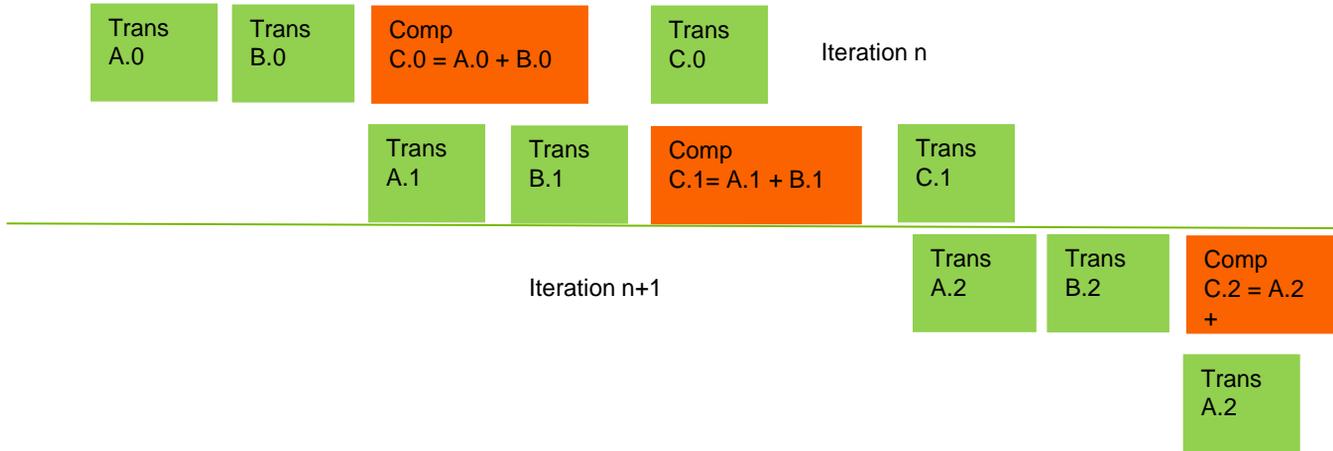
C.0 no longer blocks A.1 and B.1



Operations (Kernel launches, cudaMemcpy() calls)

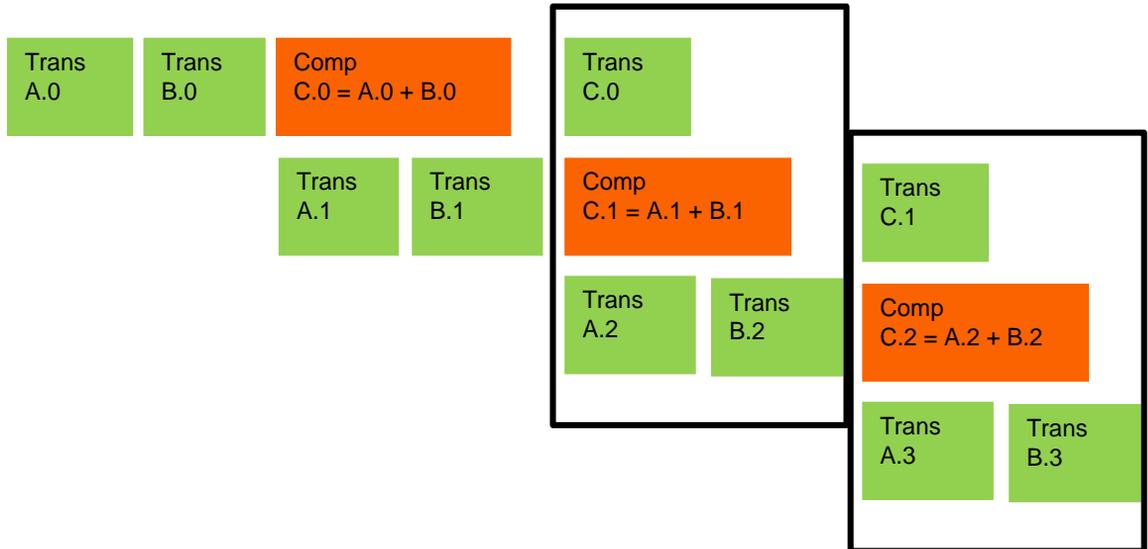
Better, not quite the best overlap

- C.1 blocks next iteration A.0 and B.0 in the copy engine queue



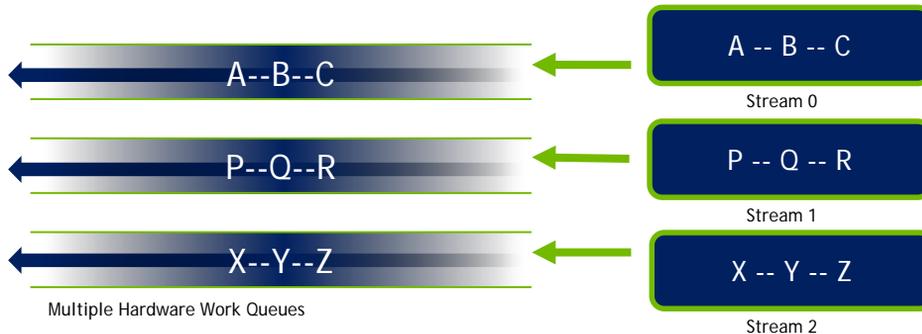
Ideal, Pipelined Timing

- Will need at least three buffers for each original A, B, and C, code is more complicated



Hyper Queues

- Provide multiple queues for each engine
- Allow more concurrency by allowing some streams to make progress for an engine while others are blocked



Wait until all tasks have completed

- `cudaStreamSynchronize(stream_id)`
 - Used in host code
 - Takes one parameter – stream identifier
 - Wait until all tasks in a stream have completed
 - E.g., `cudaStreamSynchronize(stream0)` in host code ensures that all tasks in the queues of `stream0` have completed

- This is different from `cudaDeviceSynchronize()`
 - Also used in host code
 - No parameter
 - `cudaDeviceSynchronize()` waits until all tasks in all streams have completed for the current device



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