# **LSU EE 7722**

# **Basic Setup**

Follow the instructions for class account setup on https://www.ece.lsu.edu/gp/proc.html. Code for this assignment is in directory ../hw/gpm/2024/hw02.

If the class account has been set up properly, the code can be built from within Emacs by pressing **F9** when visiting any file in the .../2024/hw02 directory or when in an Emacs shell buffer (which can be entered using **Alt**-x shell **Enter**). The code can be built from the command line using the command make -j 4 (assuming .../2024/hw02 is the current directory). Either method runs a makefile that builds all examples in the directory. It builds three versions of each program, one taking the base name of the main file, such as hw02, one with the suffix -debug, such as hw02-debug, and one with the suffix -cuda-debug, such as hw02-cuda-debug. The versions with the -cuda-debug suffix are compiled with host optimization turned off and CUDA debugging turned on, which facilitates debugging but slows down execution. To debug CUDA or host (CPU) code use the Cuda version of gdb, cuda-gdb. Note that the -cuda-debug versions will run much more slowly than the regular versions. The executables with the suffix -debug are compiled with host optimization turned off but CUDA debugging turned off. Use gdb or cuda-gdb to debug these.

Running make on a clean directory will produce a large amount of output. The make program and the file it reads, Makefile, are designed to build executables in a lazy fashion, meaning that they only create a file if it is not present or if its prerequisites have changed. Therefore a second run of make will take much less time.

Quickly check whether the build is successful with the command ./hw02 0 32. It should produce output ending with a line something like this 32 2566 6.6 27.3 3.8 8.0 4 58 1 2.

The makefile will compile code for a GPU on the system it was run. Re-run make when moving to a system using a different GPU. The Makefile should automatically detect whether the GPU for which the executable was built matches the GPU on the current system, and re-build if needed.

### **Background and Reference Material**

For this assignment one must be able to write, or at least modify, CUDA kernels. A good reference is the CUDA C++ Programming Guide, https://docs.nvidia.com/cuda/cuda-c-programming-guide/index.html. Focus on Chapter 5 up to and including 5.3 (Memory Hierarchy), but skip 5.2.1 (Thread Block Clusters). For sample code a good place to start is 2021 Homework 1, and other past assignments given in this course. The CUDA C used in this assignment is very close to C++20. A good reference for C and C++ is https://en.cppreference.com/w/.

In the references below some information is provided for specific architectures, either by CC (*e.g.*, 8.0) or by name (*e.g.*, Ampere). Both the CC 8.0 and CC 8.6 GPUs implement the Ampere architecture, 8.9 GPUs implement Ada Lovelace, and 9.0 implements Hopper. For this assignment only consider CC 8.x and 9.0 GPUs. The compute capability (CC) of the lab GPUs is shown on the system status page.

A solution to these problems requires some understanding of the hardware structure, in particular how requests are issued to the L1 cache. Some of that material is reviewed in this assignment. For additional description see Chapter 7 of the Programming Guide for the basics (but not including the L1 cache), and also Chapter 19 (Compute Capabilities) for some more details.

The hardware is covered in greater depth in the Kernel Profiling Guide,

https://docs.nvidia.com/nsight-compute/ProfilingGuide/index.html. Focus on Section 3.1 (Metrics Guide, Hardware Model) and Chapter 9 (Memory Chart). There is no need to read the material on *how* metrics are collected and there is no need to run the profiler yourself. The assignment code uses the CUPTI API to collect data. In class an MP (or SM) was described as having several-usually four-warp schedulers. The Profiling Guide refers to warp schedulers as sub partitions. For this assignment requests to the L1 cache are all global requests. Later in the semester we will make shared and maybe local requests, but probably not texture or surface requests.

# Using hw02

The code in hw02.cu contains several kernels that normalize vectors. The hw02 program takes up to three

command-line arguments: ./hw02 NBLOCKS BLOCKSIZE INPUTSIZE. The first indicates how many blocks to launch, the second indicates the number of threads per block, and the last indicates the input size.

To make sure it compiled correctly run it with arguments ./hw02 0 32, that runs the kernels fewer times (explained below). To run it while working on your solution usually run it as ./hw02.

The first argument is used to specify the number of blocks. When there are zero arguments, ./hw02, or when the first argument is zero, ./hw02 0, the number of blocks is set equal to the number of SMs. When the first argument is a positive integer, such as ./hw02 5, the kernels will be launched with that many blocks, five blocks in the example. When the first argument is a negative integer, such as ./hw02 -5, then each kernel will be launched with that many blocks *per SM*. For a GPU with 40 SMs and running with ./hw02 -5, a total of  $5 \times 40 = 200$  blocks will be launched per kernel. Note that there is no guarantee that five blocks will *simultaneously* run (be resident on) any SM, for example, if the kernels use lots of shared memory or registers fewer than five will run (and the others will have to wait).

The second argument specifies the number of warps per block. A positive value indicates the exact number of warps, for example, ./hw02 -3 4, will run each kernel with a block size of 4 warps ( $4 \times 32 = 128$  threads), and also launch 3 blocks per SM.

In many cases one wants to quickly compare the performance with different block sizes. For that omit the second argument or set it to zero, for example, ./hw02. The program will launch each kernel multiple times, starting with 1 warps per block, up to 32 warps per block. Also, because the first argument was also omitted, the number of blocks is set equal to the number of SMs. Run time and other information will be shown for each launch.

The third argument specifies the input size. If the argument is positive, is specifies the input size in MiB  $(2^{20} \text{ bytes})$ . For example, ./hw02 0 0 3.6, specifies that the input size should be 3.6 MiB. If the argument is negative then it specifies the input size in multiples of the L2 cache size. For example, ./hw02 0 0 -0.5 indicates that the input size should be half the size of the L2 cache (and so the input itself will easily fit in the L2 cache). For this assignment (Homework 1 2024) the default is  $\frac{1}{4}$  the L2 cache size, so that the input and output can both comfortably fit.

#### **Program Output**

Detailed output is obtained by running with 0 as the two command-line arguments:

### [koppel@grace hw02]\$ ./hw02 0 0

The first thing printed is information about each GPU connected to the system, followed by a line showing the chosen GPU:

GPU 0: NVIDIA GeForce RTX 4090 @ 2.52 GHz WITH 24207 MiB GLOBAL MEM GPU 0: L2: 73728 kiB MEM<->L2: 1008.1 GB/s GPU 0: CC: 8.9 SM: 128 SP-FP32/SM: 128 DP-FP64/SM: 2 TH/BL: 1024 GPU 0: SHARED: 49152 B/BL 102400 B/SM CONST: 65536 B # REGS: 65536 GPU 0: PEAK: 41288 SP GFLOPS 645 DP GFLOPS COMP/COMM: 163.8 SP 5.1 DP Using GPU 0

This assignment will only work on GPUs of CC 8 or greater.

Most fields are self-explanatory. For example, L2 is the size of the level-2 cache and CC indicates that the device is of compute capability 8.9 (Ada Lovelace). The MEM<->L2 field shows the off-chip bandwidth. SM indicates the number of streaming multiprocessors, also just called multiprocessors (MP's). CC/SM indicates the number of CUDA cores (single-precision functional units) per SM, DP/SM indicates the number of double-precision functional units per SM, and TH/BL is the maximum number of threads per block.

The amount of shared memory available is shown per block (B/BL) and per SM, this does not indicate whether any particular kernel is using that much shared memory or could use that much. The L1 cache size is usually the same size or a bit larger than the shared memory size. The same line shows the amount of constant memory, and the number of registers available.

The next line, PEAK, shows FP operation bandwidth in which a fused multiply-add is counted as one operation. (Most of the rest of the world counts a multiply-add as two operations, but in this class it's

one.) The COMP/COMM line gives the computation to communication ratio in floating-point operations per floating-point element transfers. (The information above was collected in part using the runtime library's cudaGetDeviceProperties function.)

# **Performance Data**

Each kernel is run multiple times, starting with one warp per MP, in successive runs increasing the number of warps per MP. A line of performance data is printed for each run. Appearing below is a portion of the output for an RTX 4090, showing unmodified kernel norm\_base.

```
Kernel (norm_base<4>). Uses 26 registers.
                                       n_l 1179648 d_l 4
                ----- DRAM
wp t/\mus I/el
                     N*W %pk GB/s GB/s FP \theta === Util: FP++
            BXW
                                                        Insn--
                                                               1.2** ====
                N*R
 1
    32
       5.4
            3.0
                 1.0
                     1.0
                          23 1169
                                   77
                                      292 --****
    18
2
        5.4
            3.1
                         40 2067
                 1.0
                     1.0
                                  134
                                      517 ---********
 4
    13
        5.5
            3.2
                 1.0
                     1.0
                          59 3020
                                  188
                                      755 +
            3.3
                          63 3246
8
    12
        5.6
                 1.0
                     1.0
                                  197
                                      810 +-
                                              -*********
12
    12
        5.8
            3.3
                 1.0
                     1.0
                          64 3302
                                  199
                                      816 +--
16
    14
        5.9
            3.3
                 1.0
                     1.2
                          60 3082
                                  149
                                      24
    28
        6.2
            3.2
                 1.0
                     2.5
                          45 2312
                                   85
                                      334
                         42 2193
32
    36
        6.4
            3.2
                 1.0
                     3.1
                                   65
                                      264 -*****
The lines below are fictional and are
                                   there to explain the bar graph.
    36
        6.4
            3.2
                1.0
                     3.1
                          42 2193
                                   65
                                      32
                                      32
    36
       6.4
            3.2
                1.0
                     3.1
                         42 2193
                                   65
```

The output above shows the result a kernel, norm\_base<4>. (The name shown is how the function was named, including the template parameter.)

Column wp shows the number of warps per block in the run. If the number of blocks in a launch is not set to the number of SMs then there would also be a column headed ac, which would show the number of resident warps per SM. (The number of resident warps per SM is a multiple of the number of warps per block. By default the number of blocks in a launch is set equal to the number of SMs, and in such a case the value in the ac column would match the wp column.)

For a description of the I/el, BXW, N\*R, and N\*W columns see the Base Code Performance section further below.

The columns in the L2-Cache group show how much data is moving between the L1 and L2 caches. The N\*R column (normalized amount of data read) shows how much data is read, scaled to the ideal amount. Its value is determined using a measured amount of data and a computed amount of ideal data. (Data is measured using the NVIDIA CUPTI profiling API.) A value of 1 is ideal, a value of 2 indicates that on average each element was read twice. The value under the N\*W column (normalized amount of data written) shows how much data moved from L1 to L2, normalized to the ideal amount.

The value under the GB/s in the L2 group shows the measured data throughput between the L1 and L2 caches (in either direction, but L2 to L1 dominates). The number includes all SMs. The %pk column shows this L1/L2 data movement as a percentage of peak. The DRAM GB/s column shows the measured data throughput between the L2 cache and off-chip memory.

The  $t/\mu s$  column shows the measured execution time in microseconds. To the right of  $t/\mu s$  is a bar graph showing how busy three resources are (based on certain assumptions). Three resources are tracked, FMA (fused multiply/add) instructions, shown with a +, FMA along with load instructions, shown with a -, and data transfer, shown with a \*. The data transfer shown is either SM/L2, indicated with an L2\*\* in the column heading, or L2/Mem, indicated with a Mem\*\* in the column heading. The right-most position of a resource's character indicates what fraction of the time that resource is busy. A resource is being used 100% of the time if its character reaches the rightmost position (the last = in the column heading over the bar graph).

That is true in the last line for the FMA resource, and in the penultimate line for the off-chip data transfer. In the last line we would say that the FP capability is being saturated (a good thing) and in the penultimate line we would say that data transfer is being saturated (also a good thing given the assumptions made). Those last two lines are fictional. Consider the line for the 12 warp per SM run. The \* is a bit more than halfway to the end. That indicates that L1/L2 data throughput is more than half of the peak possible. The instruction utilization, -, includes the FMAs, two loads, and one store per element.

**Problem 1:** One disappointment in the solution to Homework 1 is that when the group size is equal to the vector length many warps are required to attain good performance. For example, consider:

```
Kernel norm_group. 18 regs. D_L 8, Group sz 8, Unroll 0, n_l 589824 d_l 8
                    ----- DRAM
wp Utl t/\mus I/el
                BXW N*R
                         N*W %pk GB/s GB/s FP \theta === Util: FP++ Insn-- L2** ====
 4 1.00
        55 21.1
                 0.0
                     1.0
                         1.0
                              13
                                 692
                                       54
                                          173 -***
8 1.00
        29 21.1
                0.1
                     1.0
                         1.0
                              25 1308
                                      103
                                          327 --*****
12 1.00
        21 21.2
                0.1
                     1.0
                         1.0
                              35 1823
                                      140
                                          456 ---*******
16 1.00
        16 21.2 0.2
                     1.0
                         1.0
                             46 2351
                                      183
                                          588 ---*********
                                          24 1.00
        13 21.4
                0.2
                     1.0
                         1.0
                              56 2904
                                      225
32 1.00
        12 21.5
                0.3 1.0 1.0 60 3099
                                      243
```

The problem is that when the group size is the same as the vector size (8 in the example above) the i loops have just one iteration:

```
for ( int h = h_start; h < n_1; h += n_threads / grp_size ) {
   const size_t idx_vec_start = h * d_1;
   const size_t idx_vec_thd_start = idx_vec_start + sub_lane;
   elt_t thd_sum = 0;
   for ( int i = 0; i < d_1; i += grp_size )
      thd_sum += l_in[ idx_vec_thd_start + i ];
   const elt_t sum = group_sum(thd_sum,grp_size);
   const elt_t avg = sum / d_1;
   for ( int i = 0; i < d_1; i += grp_size )
      l_out[ idx_vec_thd_start + i ] = l_in[ idx_vec_thd_start + i ] - avg;
}</pre>
```

When the number of iterations of the i loops is large there are two advantages. Let I denote the number of iterations of the i loops. When I is large the time needed for the code outside the i loops, such as computing sum/d\_1, becomes small in comparison to the time for the i loops. (Put another way, the work needed to execute the code outside the i loop is done  $d_l/I$  times, so a larger value of I reduces the amount of work.) So, for work efficiency we want the group size to be 1, yielding the maximum  $I = d_l$ .

Recall (from class material) that for code like the *i* loops the loads will be issued first, then the computation will follow. So for the first loop if I = 4 the compiler will emit four loads, then perform the addition afterward:

```
// hw01-sol.cu:136
                           thd_sum += l_in[ idx_vec_thd_start + i ];
        /*02d0*/
                                   LDG.E.CONSTANT R4, [R2.64+-0x8] ;
        /*02e0*/
                                   LDG.E.CONSTANT R8, [R2.64+-0x4];
        /*02f0*/
                                   LDG.E.CONSTANT R10, [R2.64];
        /*0300*/
                                   LDG.E.CONSTANT R12, [R2.64+0x4];
                           thd_sum += l_in[ idx_vec_thd_start + i ];
// hw01-sol.cu:136
        /*0370*/
                                   FADD R5, RZ, R4 ;
                                   FADD R5, R5, R8;
        /*0380*/
        /*0390*/
                                   FADD R5, R5, R10;
        /*03a0*/
                                   FADD R5, R5, R12;
```

Nvidia GPUs execute LDG instructions (load global memory) by performing a lookup in the L1 cache, and if the data is not there issuing a request to the L2 cache. Execution will proceed to the next instruction

without waiting for the data to arrive. Execution only stalls (waits) for the loaded value when an attempt is made to use it. In the code above the LDG instructions will execute quickly (it will take four cycles per load in recent devices) whether or not the data is found in the cache. The first instruction to use a loaded value is the first FADD instruction, using R4 as a source, the same R4 as written by the first LDG. If the data has not yet arrived in R4 then execution will stall until the data arrives. When the data arrives the first FADD executes and execution moves to the second FADD, which references the data loaded by the second LDG carried by R8. Since the second load started four cycles after the first, the second FADD only need wait a few cycles, not the entire L1 cache miss latency.

Let *n* denote the total number of loads that need to be executed by a thread, and let *I* denote how many of those are overlapped (four in the example above), and let  $L_M$  denote load latency (say, L1 miss /L2 hit latency). Then the time needed for the loads (before data throughput approaches saturation) is  $\lceil \frac{n}{I} \rceil L_M$ . Clearly, a larger *I* is better.

In Homework 1 the norm\_group kernel was written to reduce I, thus reducing the benefits described above. Recall that in doing so other problems were reduced, including premature writebacks, bank conflicts, and cache pressure.

In this assignment the goal is to recover some of the benefits of larger *I* by properly unrolling the h loop. The standard way of unrolling a loop by *degree d* is to make *d* copies of the loop body and then simplify the *d* copies. The compiler itself won't do a good job unrolling the h loop for two reasons. First, it will assume that something done after the call to group\_sum in the original code cannot be moved before the call group\_sum. (The reason it can't move code is because group\_sum itself calls \_\_shfl\_xor\_sync, and that routine does a warp-wide barrier.) That will make it impossible to overlap loads any more than they already are. This first problem is partially solved by unrolling the h loop body in three parts: where thd\_sum is computed, where group\_sum is called, and finally where the output elements are written, each unrolled part is in a j loop:

```
for ( int hi = h_start; hi < n_l; hi += h_inc ) {</pre>
    elt_t thd_sum[unroll_degree]{};
    for ( int j = 0; j < unroll_degree; j++ )</pre>
      Ł
        int h = hi; // Homework 2: Compute h based on unroll_degree.
        const size_t idx_vec_start = h * d_l;
        const size_t idx_vec_thd_start = idx_vec_start + sub_lane;
        for ( int i = 0; i < d_1; i += grp_size )</pre>
          thd_sum[j] += l_in[ idx_vec_thd_start + i ];
      }
    elt_t avg[unroll_degree];
    for ( int j = 0; j < unroll_degree; j++ ) {</pre>
        elt_t sum = group_sum(thd_sum[j],grp_size);
        avg[j] = sum / d_1;
      }
    for ( int j = 0; j < unroll_degree; j++ ) {</pre>
        int h = hi; // Homework 2: Compute h based on unroll_degree.
        size_t idx_vec_start = h * d_l, idx_vec_thd_start = idx_vec_start + sub_lane;
        for ( int i = 0; i < d_l; i += grp_size )</pre>
          l_out[ idx_vec_thd_start + i ] = l_in[ idx_vec_thd_start + i ] - avg[j];
      }
 }
```

}

In the code above the first j loop will load unroll\_degree sets of l\_in values, and the compiler should be able to put all of those loads before the FADD instructions (because there is no intervening group\_sum).

Note that in the code above **h** is not computed properly, so though the code has the form of an unrolled loop it will actually just compute the same output vector **unroll\_loop** times.

If we wanted the compiler to unroll the h loop we could put **#pragma unroll unroll\_degree** just before the h loop. But for our h loop the compiler would not do a good job. The values of h used in the *d* copies of the body would be h, h+h\_inc, h+2\*h\_inc, ..., h+(d-1)\*h\_inc. For example, suppose n\_threads / grp\_size is 1024, unroll\_degree is 4, and h\_start is 20000. Then the values of h chosen by the compiler will be 20000, 21024, 22048, and 23072. Those values are not wrong but they are not efficient for several reasons.

First, because n\_threads is not a compile-time constant the difference between these h values is not a compile time constant and the compiler will have to emit arithmetic instructions to compute the address of each access to l\_in and l\_out. If n\_threads/grp\_size were a compile-time constant the compiler would emit instructions to compute just one address and use constant offsets.

A bigger problem with the unrolling described above is that it would bring back the problem of bank conflicts.

(a) Kernel norm\_group\_u has starter code for loop unrolling to solve the problem described above. However, as currently written the code computes the correct results but takes about unroll\_degree times as many instructions to do so. That's because the hi loop iterates the same way as the h loop in the Homework 1 solution (appearing as norm\_group in this assignment), despite the fact that the j loops do unroll\_degree times more work.

So for this assignment modify the hi loop in norm\_group\_u so that it iterates 1/unroll\_degree fewer times and also computes h correctly in the j loops.

- Avoid increasing the number of bank conflicts. That is, the number of bank conflicts for norm\_group\_u should not be higher than that of norm\_group for the same d\_l and group size.
- Compute h so that the difference between the values of h in the j loop is a compile-time constant. (This can be verified by examining the code in the SASS files.)

Output from a correctly solved assignment and an unsolved assignment appears on the following pages. The output is carefully formatted so when a page is turned from the unsolved to the solved page the corresponding output text will be in exactly the same page, which will help highlight the differences. To see this effect one must turn off continuous scrolling.

Unsolved version,  $d_l = 8$ :

```
GPU 0: NVIDIA GeForce RTX 4090 @ 2.52 GHz WITH 24088 MiB GLOBAL MEM
GPU 0: L2: 73728 kiB MEM<->L2: 1008.1 GB/s
GPU 0: CC: 8.9 SM: 128 SP-FP32/SM: 128 DP-FP64/SM: 2 TH/BL: 1024
GPU 0: SHARED: 102400 B/SM CONST: 65536 B
                                  NUM REGS: 65536
GPU 0: SHARED: 49152 B/BL SH RES: 1024 B/BL SH OPT-IN: 101376 B/BL
GPU 0: PEAK: 41288 SP GFLOPS 645 DP GFLOPS COMP/COMM: 163.8 SP 5.1 DP
Launching with 128 blocks of up to 1024 threads.
Kernel norm_group. 18 regs. D_L 8, Group sz 8, Unroll 0, n_1 589824 d_1 8
                ----- DRAM
wp Utl t/µs I/el BXW N*R N*W %pk GB/s GB/s FP \theta === Utll: FP++ Insn-- L2** ====
4 1.00 58 21.1 0.0 1.0 1.0 13 649 68 162 -***
      31 21.1 0.1 1.0 1.0 24 1225 132 306 --******
8 1.00
12 1.00
      24 21.2 0.1 1.0 1.0 31 1605 170 401 --*******
16 1.00
      17 21.2 0.2 1.0 1.0 42 2184 237 546 ---**********
      24 1.00
32 1.00
      Kernel norm_group_u. 18 regs. D_L 8, Group sz 8, Unroll 1, n_l 589824 d_l 8
                ----- DRAM
wp Utl t/μs I/el BXW N*R N*W %pk GB/s GB/s FP θ === Util: FP++ Insn-- L2** ====
4 1.00 58 21.1 0.0 1.0 1.0 13 649 70 162 -***
8 1.00 31 21.1 0.1 1.0 1.0 24 1229 132 307 --******
12 1.00 23 21.2 0.2 1.0 1.0 31 1621 173 405 --********
32 1.00
      Kernel norm_group_u. 18 regs. D_L 8, Group sz 8, Unroll 2, n_l 589824 d_l 8
                ----- DRAM
wp Utl t/μs I/el BXW N*R N*W %pk GB/s GB/s FP θ === Util: FP++ Insn-- L2** ====
4 1.00 73 31.1 0.0 1.0 1.0 10 518 56 130 -**
8 1.00 38 31.1 0.1 1.0 1.0 19 996 104 249 -****
12 1.00 28 31.2 0.1 1.0 1.0 26 1335 145 334 --******
16 1.00 21 31.2 0.2 1.0 1.0 35 1805 195 451 ---*********
Kernel norm_group_u. 16 regs. D_L 8, Group sz 8, Unroll 4, n_1 589824 d_1 8
                ----- DRAM
wp Utl t/\mus I/el BXW N*R N*W %pk GB/s GB/s FP \theta === Util: FP++ Insn-- L2** ====
4 1.00 101 49.1 0.0 1.0 1.0 7 372 40 93 -*
8 1.00 52 49.1 0.1 1.0 1.0 14 722 76 180 -****
12 1.00 38 49.2 0.1 1.0 1.0 19 986 103 246 -*****
16 1.00 28 49.2 0.2 1.0 1.0 26 1337 146 334 --******
24 1.00 21 49.3 0.3 1.0 1.0 35 1799 195 450 ---********
```

In the output above performance is worse as unroll degree increases because for degree  $\delta$  the code computes the same result  $\delta$  times. Remember this is the unsolved version, which is why the same result is computed  $\delta$  times.

Solved version,  $d_l = 8$ :

GPU 0: NVIDIA GeForce RTX 4090 @ 2.52 GHz WITH 24088 MiB GLOBAL MEM GPU 0: L2: 73728 kiB MEM<->L2: 1008.1 GB/s GPU 0: CC: 8.9 SM: 128 SP-FP32/SM: 128 DP-FP64/SM: 2 TH/BL: 1024 GPU 0: SHARED: 102400 B/SM CONST: 65536 B NUM REGS: 65536 GPU 0: SHARED: 49152 B/BL SH RES: 1024 B/BL SH OPT-IN: 101376 B/BL GPU 0: PEAK: 41288 SP GFLOPS 645 DP GFLOPS COMP/COMM: 163.8 SP 5.1 DP Launching with 128 blocks of up to 1024 threads. Kernel norm\_group. 18 regs. D\_L 8, Group sz 8, Unroll 0, n\_1 589824 d\_1 8 ----- DRAM wp Utl t/µs I/el BXW N\*R N\*W %pk GB/s GB/s FP  $\theta$  === Utll: FP++ Insn-- L2\*\* ==== 4 1.00 59 21.1 0.0 1.0 1.0 12 644 68 161 -\*\*\* 8 1.00 31 21.1 0.1 1.0 1.0 24 1231 132 308 --\*\*\*\*\*\* 12 1.00 23 21.2 0.1 1.0 1.0 31 1612 170 403 --\*\*\*\*\*\*\* 16 1.00 17 21.2 0.2 1.0 1.0 42 2189 236 547 ---\*\*\*\*\*\*\*\*\*\*\* 24 1.00 32 1.00 Kernel norm\_group\_u. 18 regs. D\_L 8, Group sz 8, Unroll 1, n\_l 589824 d\_l 8 ----- DRAM wp Utl t/μs I/el BXW N\*R N\*W %pk GB/s GB/s FP θ === Util: FP++ Insn-- L2\*\* ==== 4 1.00 59 23.1 0.0 1.0 1.0 12 635 68 159 -\*\*\* 8 1.00 31 23.1 0.1 1.0 1.0 23 1204 131 301 --\*\*\*\*\* 12 1.00 24 23.2 0.1 1.0 1.0 31 1604 170 401 --\*\*\*\*\*\*\*\* 32 1.00 Kernel norm\_group\_u. 20 regs. D\_L 8, Group sz 8, Unroll 2, n\_l 589824 d\_l 8 ----- DRAM wp Utl t/μs I/el BXW N\*R N\*W %pk GB/s GB/s FP θ === Util: FP++ Insn-- L2\*\* ==== 4 1.00 33 16.6 0.0 1.0 1.0 22 1138 126 284 --\*\*\*\*\* 8 1.00 19 16.6 0.2 1.0 1.0 39 2024 218 506 ---\*\*\*\*\*\*\*\*\*\*\* Kernel norm\_group\_u. 26 regs. D\_L 8, Group sz 8, Unroll 4, n\_l 589824 d\_l 8 ----- DRAM wp Utl t/µs I/el BXW N\*R N\*W %pk GB/s GB/s FP  $\theta$  === Utll: FP++ Insn-- L2\*\* ==== 4 1.00 20 13.3 0.1 1.0 1.0 36 1874 206 468 ---\*\*\*\*\*\*\*\* Kernel norm\_group\_u. 33 regs. D\_L 8, Group sz 8, Unroll 8, n\_l 589824 d\_l 8 ----- DRAM wp Utl t/µs I/el BXW N\*R N\*W %pk GB/s GB/s FP  $\theta$  === Util: FP++ Insn-- L2\*\* ==== 8 1.00 12 1.00 16 1.00 32 0.90 

In the output pay attention to the number of instructions per iteration (I/el) and the number of bank conflicts BXW. Output from the unsolved version appeared on the previous page.

Unsolved version,  $d_l = 32$ :

GPU 0: NVIDIA GeForce RTX 4090 @ 2.52 GHz WITH 24088 MiB GLOBAL MEM GPU 0: L2: 73728 kiB MEM<->L2: 1008.1 GB/s GPU 0: CC: 8.9 SM: 128 SP-FP32/SM: 128 DP-FP64/SM: 2 TH/BL: 1024 GPU 0: SHARED: 102400 B/SM CONST: 65536 B NUM REGS: 65536 GPU 0: SHARED: 49152 B/BL SH RES: 1024 B/BL SH OPT-IN: 101376 B/BL GPU 0: PEAK: 41288 SP GFLOPS 645 DP GFLOPS COMP/COMM: 163.8 SP 5.1 DP Launching with 128 blocks of up to 1024 threads. Kernel norm\_group. 16 regs. D\_L 32, Group sz 32, Unroll 0, n\_l 147456 d\_l 32 ----- DRAM wp Utl t/µs I/el BXW N\*R N\*W %pk GB/s GB/s FP  $\theta$  === Utll: FP++ Insn-- L2\*\* ==== 4 1.00 66 26.1 0.0 1.0 1.0 11 576 62 144 -\*\*\* 8 1.00 35 26.1 0.1 1.0 1.0 21 1088 117 272 --\*\*\*\*\* 12 1.00 26 26.2 0.2 1.0 1.0 28 1458 158 365 --\*\*\*\*\*\* Kernel norm\_group\_u. 16 regs. D\_L 32, Group sz 32, Unroll 1, n\_l 147456 d\_l 32 ----- DRAM wp Utl t/μs I/el BXW N\*R N\*W %pk GB/s GB/s FP θ === Util: FP++ Insn-- L2\*\* ==== 4 1.00 66 26.1 0.0 1.0 1.0 11 573 61 143 -\*\*\* 8 1.00 34 26.1 0.1 1.0 1.0 21 1102 116 275 --\*\*\*\*\* 12 1.00 26 26.2 0.2 1.0 1.0 28 1439 154 360 --\*\*\*\*\*\* 24 1.00 15 26.4 0.3 1.0 1.0 49 2541 275 635 +---\*\*\*\*\*\*\*\*\*\*\* Kernel norm\_group\_u. 20 regs. D\_L 32, Group sz 32, Unroll 2, n\_l 147456 d\_l 32 ----- DRAM wp Utl t/µs I/el BXW N\*R N\*W %pk GB/s GB/s FP  $\theta$  === Utll: FP++ Insn-- L2\*\* ==== 4 1.00 81 36.1 0.0 1.0 1.0 9 467 50 117 -\*\* 8 1.00 42 36.1 0.1 1.0 1.0 17 902 94 226 -\*\*\*\* 12 1.00 31 36.2 0.2 1.0 1.0 23 1207 130 302 --\*\*\*\*\*\* 16 1.00 23 36.2 0.3 1.0 1.0 32 1628 178 407 --\*\*\*\*\*\*\* 24 1.00 18 36.3 0.4 1.0 1.0 42 2153 232 538 ---\*\*\*\*\*\*\*\*\*\* Kernel norm\_group\_u. 15 regs. D\_L 32, Group sz 32, Unroll 4, n\_l 147456 d\_l 32 ----- DRAM wp Utl t/ $\mu$ s I/el BXW N\*R N\*W %pk GB/s GB/s FP  $\theta$  === Util: FP++ Insn-- L2\*\* ==== 4 1.00 111 56.1 0.0 1.0 1.0 7 341 37 85 \*\* 8 1.00 57 56.1 0.1 1.0 1.0 13 662 68 165 -\*\*\* 12 1.00 41 56.2 0.2 1.0 1.0 18 915 95 229 -\*\*\*\*\* 16 1.00 31 56.2 0.3 1.0 1.0 24 1227 130 307 --\*\*\*\*\*\* 24 1.00 23 56.3 0.4 1.0 1.0 31 1615 174 404 --\*\*\*\*\*\*\* 32 1.00 21 56.4 0.5 1.0 1.0 35 1821 200 455 ---\*\*\*\*\*\*\*\*\*

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Solved version,  $d_l = 32$ :

GPU 0: NVIDIA GeForce RTX 4090 @ 2.52 GHz WITH 24088 MiB GLOBAL MEM GPU 0: L2: 73728 kiB MEM<->L2: 1008.1 GB/s GPU 0: CC: 8.9 SM: 128 SP-FP32/SM: 128 DP-FP64/SM: 2 TH/BL: 1024 GPU 0: SHARED: 102400 B/SM CONST: 65536 B NUM REGS: 65536 GPU 0: SHARED: 49152 B/BL SH RES: 1024 B/BL SH OPT-IN: 101376 B/BL GPU 0: PEAK: 41288 SP GFLOPS 645 DP GFLOPS COMP/COMM: 163.8 SP 5.1 DP Launching with 128 blocks of up to 1024 threads. Kernel norm\_group. 16 regs. D\_L 32, Group sz 32, Unroll 0, n\_l 147456 d\_l 32 ----- DRAM wp Utl t/µs I/el BXW N\*R N\*W %pk GB/s GB/s FP  $\theta$  === Utll: FP++ Insn-- L2\*\* ==== 4 1.00 66 26.1 0.0 1.0 1.0 11 575 62 144 -\*\*\* 8 1.00 35 26.1 0.1 1.0 1.0 21 1089 115 272 --\*\*\*\*\* 12 1.00 26 26.2 0.2 1.0 1.0 28 1458 156 364 --\*\*\*\*\*\* Kernel norm\_group\_u. 16 regs. D\_L 32, Group sz 32, Unroll 1, n\_l 147456 d\_l 32 ----- DRAM wp Utl t/μs I/el BXW N\*R N\*W %pk GB/s GB/s FP θ === Util: FP++ Insn-- L2\*\* ==== 4 1.00 65 26.0 0.0 1.0 1.0 11 577 62 144 -\*\*\* 8 1.00 34 26.1 0.1 1.0 1.0 21 1103 118 276 --\*\*\*\*\* 12 1.00 26 26.1 0.2 1.0 1.0 28 1468 156 367 --\*\*\*\*\*\* 24 1.00 15 26.3 0.3 1.0 1.0 49 2528 272 632 +---\*\*\*\*\*\*\*\*\*\*\* Kernel norm\_group\_u. 18 regs. D\_L 32, Group sz 32, Unroll 2, n\_l 147456 d\_l 32 ----- DRAM wp Utl t/µs I/el BXW N\*R N\*W %pk GB/s GB/s FP  $\theta$  === Util: FP++ Insn-- L2\*\* ==== 4 1.00 40 20.1 0.1 1.0 1.0 18 952 178 238 -\*\*\*\* 8 1.00 20 20.1 0.2 1.0 1.0 36 1859 202 465 ---\*\*\*\*\*\*\*\*\* Kernel norm\_group\_u. 26 regs. D\_L 32, Group sz 32, Unroll 4, n\_l 147456 d\_l 32 ----- DRAM wp Utl t/µs I/el BXW N\*R N\*W %pk GB/s GB/s FP  $\theta$  === Utll: FP++ Insn-- L2\*\* ==== 4 1.00 22 17.1 0.2 1.0 1.0 34 1746 194 436 --\*\*\*\*\*\*\*\* Kernel norm\_group\_u. 35 regs. D\_L 32, Group sz 32, Unroll 8, n\_l 147456 d\_l 32 ----- DRAM wp Utl t/μs I/el BXW N\*R N\*W %pk GB/s GB/s FP θ === Util: FP++ Insn-- L2\*\* ====