CUDA
MATMULT
REG / SHARED

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SHARED/SHARED OPERANDS

In-product in Matmult PA1:

for(int e=0; e<BLOCK_SIZE; ++e)
    Cvalue += shared_A[thread_row][e] * shared_B[e][thread_col];

Cuda allows both operands of operations (e.g. multiply add) to be in shared memory.

But, is that efficient?
MICRO BENCHMARK

Do a micro benchmark.

Like we studied coalescing in vector add micro benchmark.

Micro benchmark: comparison of 2 (or more) ways of performing a computation in a simplest possible code, that only changes ONE aspect of that code, with the goal of understanding that one aspect.

In our case here: what is the cost, if any, of having both operands in shared memory vs the cost of 1 operand in a register and 1 in shared memory.

Once we understand this behavior we can exploit it in a larger application, e.g. matmult
// sha sha
for(i = 0; i < NUMREPS; i++)
    for(j = 0; j < VALSPERTHREAD; j++)
        local_C += shared_A[j] * shared_B[j];

// reg sha
for(i = 0; i < NUMREPS; i++)
    for(j = 0; j < VALSPERTHREAD; j++)
        local_C += shared_A[j] * local_B[j];

// And that is the ONLY difference in the whole code!!
SHARED SHARED VS REG SHARED ASM

label1:

mov.b32 $r0, s[0x0060]
mad.rn.f32 $r1, s[0x0020], $r0, $r1
mov.b32 $r0, s[0x0064]
mad.rn.f32 $r1, s[0x0024], $r0, $r1

... 
mov.b32 $r0, s[0x009c]
mad.rn.f32 $r1, s[0x005c], $r0, $r1
add.b32 $r31, $r31, 0x00000001
set.ne.s32 $p0|$o127, $r31, c1[0x0000]
@$p0.ne bra.label label1

shared shared     twice as slow as     reg shared

label1:
mad.rn.f32 $r3, s[0x0020], $r2, $r3
mad.rn.f32 $r3, s[0x0024], $r4, $r3

... 
mad.rn.f32 $r3, s[0x005c], $r32, $r3
add.b32 $r0, $r0, 0x00000001
set.ne.s32 $p0|$o127, $r0, c1[0x0000]
@$p0.ne bra.label label1
Thread block: 64x1, C footprint: 64x16

- each thread has a column of 16 C values in registers
- dim3 dimBlock(64,1)
- dim3 dimGrid(C.width/64,C.height/16)

A is read into shared memory in 16x16 blocks

B is read into registers (1 per thread) one row part at the time

better performance: pull loop invariant code out
for (b=0; b<A.width/16; ++b)
    // locate A block
    // read 4 A elements into shared memory As
    // locate base: starting row part in B
    for (k=0; k<16; ++k)
        // read B element Bel ( B[k,j] )
        for (i=0; i<16; ++i)
            C_local[i] += As[i*16+k]*Bel

for (i=0; i<16; ++i)
    write C_local[i] to C
PERFORMANCE

We compare our code to the matrix multiply code from CudaBLAS for square 2048x2048 matrices on a Tesla1060. Our original 16x64 C footprint register shared code:

- 265 GFLOPS (GigaFlops per Second)

CudaBLAS

- 354 GFLOPS
PERFORMANCE

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First optimization: loop unrolling, pointer arithmetic (avoiding a*i+b by pointer addition), code hoisting (taking loop invariant code out of loop bodies) 321 GFLOPS
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Second optimization

32x64 C footprint (less glob. mem. traffic)

4x16 thread block (simpler mapping of thread space to data space)

Transposing 32x16 A into 16x33 A-shared block

(accesses are consecutive, no bank conflicts) 372 GFLOPS
CAN WE DO BETTER?

• REG/REG: Every operand in register?

• We never made it work, but maybe you can (we have bigger GPUs now with more registers....)