Lab 4: Multicore Parallelism

After this lab, you'll be able to...

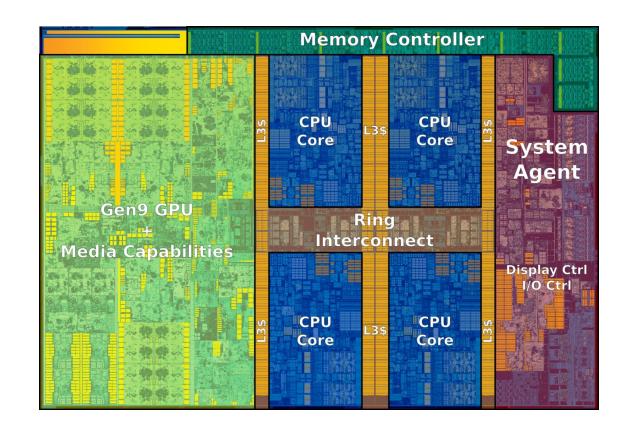
- Understand the basic training algorithm our CNN framework uses
- Optimize the performance of loops using OpenMP
- Measure the impact of your optimizations
- Understand how convolution, pool, and RELU layers work in a CNN.



Multiple Processors

- We have four cores, but we've only been using one!
- We can share work across the cores to go faster and/or for lower energy.

 Take CSE 160 for more extensive treatment



Concurrency

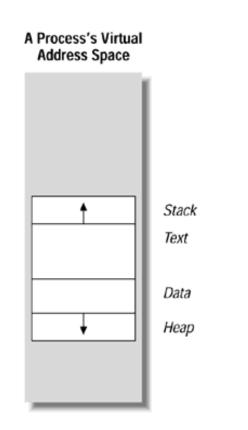
- Different pieces of work are being done by different threads
- If they are running at the same time, we call this parallel execution.

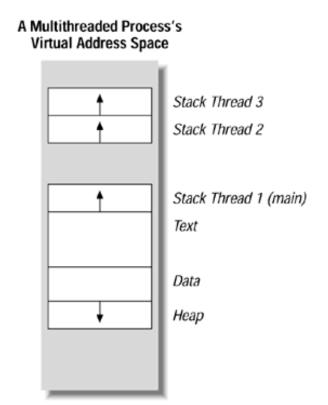
Multi-processing vs. multi-threading



Threads

- Shared Memory Model
 - All threads see the same view of memory (except locals)
 - Each thread has its own stack space and PC
- OS runs threads in any order
- For correctness you may need to use synchronization







How To Use Threads (CSE 141L)

- Option 1: OpenMP (Open multiprocessing)
 - High-level API for parallelizing C and C++ code
 - Relatively easy
 - Only applicable to relatively simple loops
- Option 2: pthreads
 - Low-level API
 - Harder to use
 - You can use it for many more things
 - Take CSE120



Hello World in openmp

- Need to use –fopen and <omp.h>
- Options all start with "#pragma omp"
 - "#pragma ..." is a way to add arbitrary extension to C/C++
- "#pragma omp parallel"
 - Means give me a bunch of threads (system default)
 - compiler uses pthreads under the hood
- "omp_get_thread_num()"
 - Gets you the number of your thread (0...n)



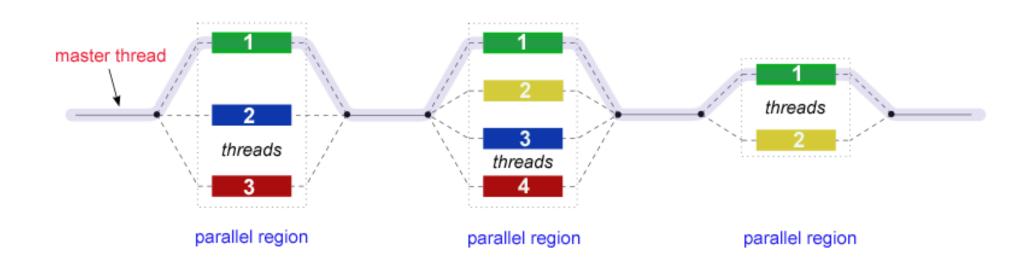
Synchronization

- Things can go terribly wrong working with shared variables
 - Race conditions output depends on which thread comes first
 - BUT, your program must be correct for all orderings
- We can avoid this by using synchronization
 - E.g., #pragma omp critical
 - Under the hood, these are done with mutexes or locks (CSE120)
 - But synchronization means serialization
 - Do it too much, you lose all performance benefits



Fork-Join Parallelism

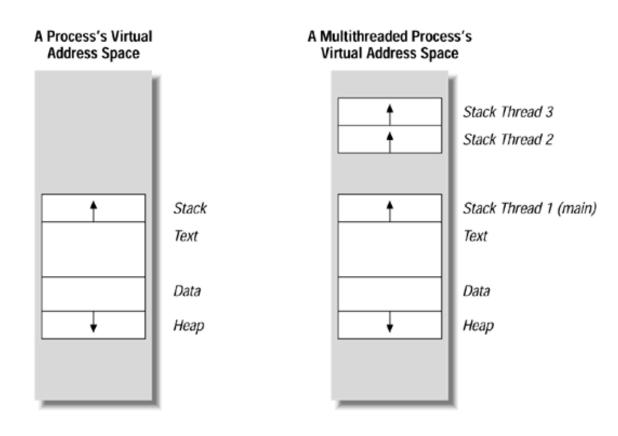
• Do some work in serial, do a bunch in parallel, join the results, repeat.



Managing Data

- Data allocated outside a pragma is shared
- Data allocated inside a pragma is private

 Be careful – this is absolutely essential





Calculating pi

Classic approach, solve:

$$\int_{0}^{1} \frac{4.0}{(1+x^2)} dx = \pi$$

 This has some really nice loop independence we can exploit

```
static long num_steps = 100000;
double step;
int main ()
{
  int i; double x, pi, sum = 0.0;
  step = 1.0/(double) num_steps;
  for (i=0;i< num_steps; i++){
      x = (i+0.5)*step;
      sum = sum + 4.0/(1.0+x*x);
  }
  pi = step * sum;
}</pre>
```

Calculating pi – in parallel

- From openmp, only use:
 - #pragma omp parallel, omp_get_num_threads(), omp_get_thread_num()
 - Careful declaration of local vs. global vars
 - (Hint, an array can be global but each thread only need access each part)



Synchronization Revisited

- Barriers
 - Automatic at the end of a parallel pragma, but can also do:
 - "#pragma omp barrier" within a pragma
- Mutual Exclusion
 - #pragma omp critical



Synchronization Revisited

- Barriers
 - Automatic at the end of a parallel pragma, but can also do:
 - "#pragma omp barrier" within a pragma
- Mutual Exclusion
 - #pragma omp critical
 - #pragma omp atomic
 - Only applies to a single statement, can be faster than omp critical by leveraging some atomic hardware instructions



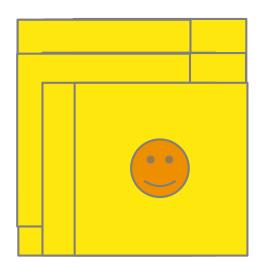
OpenMP Primitives: Loops

- #pragma omp parallel for
 - Run the following for loop with multiple threads.
 - The loop needs to be pretty simple.
 - Something like "for(int i = C; i < K; i+=B)"</p>
 - K and B need to fixed for the execution of the loop
 - Otherwise, nothing will happen.
 - It uses all the cores by default.



Image Alignment

- We are going to do image alignment on a batch of images (i.e., frames of video)
 - Images are 228x228 (x = y = 228) gray scale (z = 1)
 - For frame b, we will shift around frame b 1
 - Compute the sum of absolute differences for each offset.
 - Offset will be 0-7 for and 0-7 for y
 - Store the result in an output tensor
 - X = 8 (offsets)
 - Y = 8 (offsets)
 - Z = 1
 - B = B (there will be one extra output)



X-offset

Y-offset	1	1	1
	1	0.1	1
<i></i>	1	1	1



Tiled Example Code

```
void do_stabilize_tile_y_1(const tensor_t<double> & images, tensor_t<double> & output, int TILE_SIZE)
        for (int this_frame = 1; this_frame < images.size.b; this_frame++) {</pre>
                int previous_frame = this_frame - 1;
                for(int pixel_yy = 0; pixel_yy < images.size.y; pixel_yy += TILE_SIZE) {</pre>
                         for (int offset_x = 0; offset_x < MAX_OFFSET; offset_x++) {</pre>
                                 for (int offset_v = 0; offset_v < MAX_OFFSET; offset_v++) {</pre>
                                         for(int pixel_y = pixel_yy; pixel_y < pixel_yy + TILE_SIZE && pixel_y < images.size.y; pixel_y++) {</pre>
                                                  for(int pixel_x = 0; pixel_x < images.size.x; pixel_x++) {</pre>
                                                          int shifted x = pixel x + offset x;
                                                          int shifted_y = pixel_y + offset_y;
                                                          if (shifted_x >= images.size.x ||
                                                              shifted_y >= images.size.y)
                                                                  continue:
                                                          output(offset_x, offset_y, 0, this_frame) +=
                                                                  fabs(images(pixel_x, pixel_y, 0, this_frame) -
                                                                        images(shifted_x, shifted_y, 0, previous_frame));
```

- Iterations of the outer loop don't write-share anything
 - Each iteration writes to output(..., this_frame) and nothing else
 - They can run in parallel
- Execution time: 0.516s

Parallelized Version

```
void do_stabilize_tile_y_1_omp_simple(const tensor_t<double> & images, tensor_t<double> & output, int TILE_SIZE)
        OPEN_TRACE("trace.out");
        // parallizing across results in no sharing in `output` since
        // each frame output it's result to one element of `output`.
#pragma omp parallel for
            (int this_frame = 1; this_frame < images.size.b; this_frame++) {</pre>
                int previous frame = this frame - 1;
                for(int pixel_yy = 0; pixel_yy < images.size.y; pixel_yy += TILE_SIZE) {</pre>
                        for (int offset_x = 0; offset_x < MAX_OFFSET; offset_x++) {</pre>
                                 for (int offset_y = 0; offset_y < MAX_OFFSET; offset_y++) {</pre>
                                         for(int pixel_y = pixel_yy; pixel_y < pixel_yy + TILE_SIZE && pixel_y < images.size.y; pixel_y++) {</pre>
                                                 for(int pixel_x = 0; pixel_x < images.size.x; pixel_x++) {</pre>
                                                          int shifted_x = pixel_x + offset_x;
                                                          int shifted_y = pixel_y + offset_y;
                                                          if (shifted_x >= images.size.x ||
                                                              shifted_y >= images.size.y)
                                                                  continue;
                                                          //DUMP ACCESSES();
                                                          double t = fabs(images(pixel_x, pixel_y, 0, this_frame) -
                                                                          images(shifted_x, shifted_y, 0, previous_frame));
                                                          output(offset_x, offset_y, 0, this_frame) += t;
```

- The #pragma parallelizes the next loop.
- Iterations of that loop will run in parallel.
 - Our tiling and loop re-nesting optimizations work as they did before.
- Execution time: 0.173 (2.99x)

Another Approach

```
void do_stabilize_tile_y_1(const tensor_t<double> & images, tensor_t<double> & output, int TILE_SIZE)
        for (int this_frame = 1; this_frame < images.size.b; this_frame++) {</pre>
                int previous_frame = this_frame - 1;
                for(int pixel_yy = 0; pixel_yy < images.size.y; pixel_yy += TILE_SIZE) {</pre>
                         for (int offset_x = 0; offset_x < MAX_OFFSET; offset_x++) {</pre>
                                 for (int offset_v = 0; offset_v < MAX_OFFSET; offset_v++) {</pre>
                                         for(int pixel_y = pixel_yy; pixel_y < pixel_yy + TILE_SIZE && pixel_y < images.size.y; pixel_y++) {</pre>
                                                  for(int pixel_x = 0; pixel_x < images.size.x; pixel_x++) {</pre>
                                                          int shifted x = pixel x + offset x;
                                                          int shifted_y = pixel_y + offset_y;
                                                          if (shifted_x >= images.size.x ||
                                                              shifted_y >= images.size.y)
                                                                  continue:
                                                          output(offset_x, offset_y, 0, this_frame) +=
                                                                  fabs(images(pixel_x, pixel_y, 0, this_frame) -
                                                                        images(shifted_x, shifted_y, 0, previous_frame));
```

- Let's try parallelizing pixel_yy instead.
- It might put less pressure on the L3 since we only need to have 2 frames in memory instead of images.size.b

```
Another Approach:
void do_stabilize_tile_y_1_omp_critical(const tensor_t<doubl
         //OPEN_TRACE("trace.out");
         for (int this_frame = 1; this_frame < images.size.b;</pre>
                  int previous_frame = this_frame - 1;
                                                                                 128
                                                                                                   113
                                                                          91.7
                                                                                 134
                                                                                       128
                                                                                             112
                                                                                                  91.4
                                                                                                         87.7
                                                                          95.4
                                                                                 137
                                                                                       108
                                                                                            96.1
                                                                                                  69.8
                                                                                                               103
                  77 Parallelizing on pixel_yy, results in sha
                                                                          93.5
                                                                                      94.9
                                                                                            76.7
                                                                                                               110
                                                                                                                      122
                                                                                                   58.7
                                                                                            61.3
                                                                                                  60.2
#pragma omp parallel for
                                                                           136
                                                                                      83.7
                                                                                                         119
                                                                                                                      130
                                                                                                                      126
                            pixel_vy = 0; pixel_vy < images.size</pre>
                            for (int offset x = 0; offset x < MAX
                                     for (int offset_y = 0; offse
                                                                                                                                                       ixel v++)
                                               for(int pixel v = pi
                                                        for(int pixe
                                                                        -40 -4.8 -18 -17 -28 -10 -18 -23
                                                                        -14 -6.8 -15 -28 -41 -13 -16 -18
                                                                       stabilize.exe: build/stabilize.cpp:532: void stabilize(const string&, const dataset_t&, int): Assertion `O' failed
                                                                       Aborted (core dumped)
                                                                                     images(snirted_x, snirted_y, 0, previous_frame));
                                                                 output(offset_x, offset_y, 0, this_frame) += t;
```

- Just move the #pragma
- Execution time: ???

The threads are working on different "slices" of pixel_yy, but they are updating same output array.

Critical section!

- Because multiple threads are working with the same data at the same time, we need:
 - #pragma omp critical



Another Approach: Parallelize inner loop

```
void do stabilize tile y 1 omp critical(const tensor t<double> & images, tensor t<double> & output, int TILE SIZE)
        OPEN_TRACE("trace.out");
        for (int this_frame = 1; this_frame < images.size.b; this_frame++) {</pre>
                int previous_frame = this_frame - 1;
                // Parallelizing on pixel yy, results in sharing in output.
#pragma omp parallel for
                for(int pixel_yy = 0; pixel_yy < images.size.y; pixel_yy += TILE_SIZE) {</pre>
                         for (int offset_x = 0; offset_x < MAX_OFFSET; offset_x++) {</pre>
                                 for (int offset_y = 0; offset_y < MAX_OFFSET; offset_y++) {</pre>
                                          for(int pixel_y = pixel_yy; pixel_y < pixel_yy + TILE_SIZE && pixel_y < images.size.y; pixel_y++) {</pre>
                                                  for(int pixel_x = 0; pixel_x < images.size.x; pixel_x++) {</pre>
                                                          int shifted_x = pixel_x + offset_x;
                                                          int shifted_y = pixel_y + offset_y;
                                                          if (shifted_x >= images.size.x ||
                                                              shifted_y >= images.size.y)
                                                                   continue:
                                                          double t = fabs(images(pixel_x, pixel_y, 0, this_frame) -
                                                                           images(shifted_x, shifted_y, 0, previous_frame));
                                                           #pragma omp critical
                                                                  output(offset_x, offset_y, 0, this_frame) += t;
```

- This will ensure that the threads access output one at a time.
- Execution time: 23s (0.02x speedup)
- Critical sections incur overhead, and we are using too many of them.

```
void do_stabilize_tile_y_1_omp_critical_fast(const tensor_t<double> & images, tensor_t<double> & output, int TILE_SIZE)
        OPEN_TRACE("trace.out");
        for (int this_frame = 1; this_frame < images.size.b; this_frame++) {</pre>
                int previous_frame = this_frame - 1;
                // same thing: need to protect
#pragma omp parallel for
                for(int pixel_yy = 0; pixel_yy < images.size.y; pixel_yy += TILE_SIZE) {</pre>
                        tensor_t<double>_output(output.size);
                        _output.clear();
                        for (int offset_x = 0; offset_x < MAX_OFFSET; offset_x++) {</pre>
                                for (int offset_y = 0; offset_y < MAX_OFFSET; offset_y++) {</pre>
                                         for(int pixel_y = pixel_yy; pixel_y < pixel_yy + TILE_SIZE && pixel_y < images.size.y; pixel_y++) {</pre>
                                                 for(int pixel_x = 0; pixel_x < images.size.x; pixel_x++) {</pre>
                                                         int shifted_x = pixel_x + offset_x;
                                                         int shifted_y = pixel_y + offset_y;
                                                         if (shifted_x >= images.size.x ||
                                                             shifted_y >= images.size.y)
                                                                 continue;
                                                         double t = fabs(images(pixel_x, pixel_y, 0, this_frame) -
                                                                         images(shifted_x, shifted_y, 0, previous_frame));
                                                         // accumulate the updates locally
                                                         _output(offset_x, offset_y, 0, this_frame) += t;
                                         }
#pragma omp critical // Apply them en masse this is reasonably fast because it's small, so the serialization isn't a big deal.
                                for (int offset_y = 0; offset_y < MAX_OFFSET; offset_y++) {</pre>
                                        for (int offset_x = 0; offset_x < MAX_OFFSET; offset_x++) {</pre>
                                                 output(offset_x, offset_y, 0, this_frame) += _output(offset_x, offset_y, 0, this_frame);
                                                 output is local to each thread, no critical section needed
                                                 At the end of each iteration, update outputs all at once.

    Fewer critical sections

                                                  Execution time: 0.115 (4.48x)
```

What do you need to use?

- Just these two primitives are enough for our solution:
 - #pragma omp parallel for
 - #pragma omp critical
- But you'll need to be careful where to put these.
- <Warning> gprof doesn't work on multithreaded programs



There's a Lot More in OpenMP

- You don't need it for the lab, but you are welcome use it.
- There are lots of documents online.
 - Many of them are not very useful because they are too detailed
- This blog is pretty good
 - http://jakascorner.com/blog/
 - Especially these
 - http://jakascorner.com/blog/2016/04/omp-introduction.html
 - http://jakascorner.com/blog/2016/05/omp-for.html
 - http://jakascorner.com/blog/2016/06/omp-for-scheduling.html
 - http://jakascorner.com/blog/2016/06/omp-data-sharing-attributes.html
 - http://jakascorner.com/blog/2016/07/omp-default-none-and-const.html



CSE141L - Lab 4 - Threads

Discussion Session

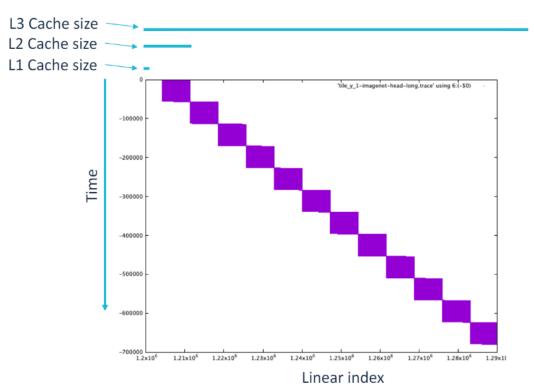
Overview

- Questions about multithreading in stabilize
- Understanding the Memory Access Visualization Tool
- Walkthrough of OpenMP on fc_layer_t::calc_grads()
 - #pragma omp parallel for
 - #pragma omp critical
- Brief overview of fc_layer_t::fix_weights
- Questions

Questions about multithreading in stabilize?

```
pragma omp parallel for
for(int pixel_yy = 0; pixel_yy < images.size.y; pixel_yy += TILE_SIZE) {</pre>
   tensor_t<double> output(output.size);
   output.clear();
   for (int offset x = 0; offset x < MAX OFFSET; offset x++)
       for (int offset y = 0; offset y < MAX OFFSET; offset y++) {</pre>
           for(int pixel y = pixel yy; pixel y < pixel yy + TILE SIZE && pixel y < images.size.y; pixel y++) {</pre>
               for(int pixel_x = 0; pixel_x < images.size.x; pixel_x++) {</pre>
                   int shifted x = pixel x + offset x;
                   int shifted v = pixel v + offset v;
                   if (shifted x >= images.size.x ||
                       shifted y >= images.size.y)
                    double t = fabs(images(pixel x, pixel y, 0, this frame) -
                           images(shifted x, shifted y, 0, previous frame));
                   // accumulate the updates locally
                   output(offset x, offset y, 0, this frame) += t;
   #pragma omp critical // Apply them en masse this is reasonably fast because it's small, so the serialization isn't a big deal.
       for (int offset y = 0; offset y < MAX OFFSET; offset y++) {</pre>
           for (int offset x = 0; offset x < MAX OFFSET; offset x++) {</pre>
               output(offset x, offset y, 0, this frame) += output(offset x, offset y, 0, this frame);
```

Memory Access Visualization Tool



Original - fc_layer_t::calc_grads

```
void calc_grads( const tensor_t<double>& grad_next_layer ) {
   memset( grads_out.data, 0, grads_out.size.x * grads_out.size.y * grads_out.size.z * sizeof( double ) );
    grads out.size.x = grads out.size.x * grads out.size.y * grads out.size.z;
    grads out.size.y = 1;
    grads out.size.z = 1:
    for ( int b = 0; b < out.size.b; b++ ) {
        for ( int n = 0; n < activator_input.size.x; n++ ){</pre>
            double ad = activator_derivative( activator_input(n, 0, 0, b) );
            double ng = grad_next_layer(n, 0, 0, b);
            act_grad(n, 0, 0, b) = ad * ng;
    for ( int b = 0; b < out.size.b; b++ ) {
        for ( int n = 0; n < weights.size.y; n++ ) {</pre>
            for ( int i = 0; i < weights.size.x; i++ ) {
                    grads_out(i, 0, 0, b) += act_grad(n, 0, 0, b) * weights( i, n, 0);
    grads out.size = in.size:
```

Original - fc_layer_t::calc_grads

```
void calc_grads( const tensor_t<double>& grad_next_layer ) {
   memset( grads_out.data, 0, grads_out.size.x * grads_out.size.y * grads_out.size.z * sizeof( double ) );
    grads_out.size.x = grads_out.size.x * grads_out.size.y * grads_out.size.z;
    grads out.size.y = 1;
    grads_out.size.z = 1;
    for ( int b = 0; b < out.size.b; b++ ) {
                                                                                       Calculate derivative
        for ( int n = 0; n < activator_input.size.x; n++ ){</pre>
                                                                                       of activator function
            double ad = activator_derivative( activator_input(n, 0, 0, b) );
            double ng = grad_next_layer(n, 0, 0, b);
                                                                                       Backpropagate Error
            act_grad(n, 0, 0, b) = ad * ng;
                                                                                       from Next Layer
    for ( int b = 0; b < out.size.b; b++ ) {
        for ( int n = 0; n < weights.size.y; n++ ) {</pre>
            for ( int i = 0; i < weights.size.x; i++ ) {
                    grads_out(i, 0, 0, b) += act_grad(n, 0, 0, b) * weights(i, n, 0);
    grads_out.size = in.size;
```

Original - fc_layer_t::calc_grads

```
void calc_grads( const tensor_t<double>& grad_next_layer ) {
   memset( grads_out.data, 0, grads_out.size.x * grads_out.size.y * grads_out.size.z * sizeof( double ) );
    grads_out.size.x = grads_out.size.x * grads_out.size.y * grads_out.size.z;
    grads out.size.y = 1;
    grads out.size.z = 1:
    for ( int b = 0; b < out.size.b; b++ ) {
        for ( int n = 0; n < activator_input.size.x; n++ ){</pre>
            double ad = activator_derivative( activator_input(n, 0, 0, b) );
            double ng = grad_next_layer(n, 0, 0, b);
            act_grad(n, 0, 0, b) = ad * ng;
    for ( int b = 0; b < out.size.b; b++ ) {
        for ( int n = 0; n < weights.size.y; n++ ) {</pre>
                                                                                             Find error of each
            for ( int i = 0; i < weights.size.x; i++ ) {
                                                                                             input proportional to
                    grads_out(i, 0, 0, b) += act_grad(n, 0, 0, b) * weights(i, n, 0);
                                                                                             its weight and sum it
                                                                                             up
    grads_out.size = in.size;
```

Baseline - fc_calc_grads (Tier 1)

```
void calc grads( const tensor t<double>& grad next layer ) {
    memset( grads out.data, 0, grads out.size.x * grads out.size.y * grads out.size.z * sizeof( double ) );
    grads out.size.x = grads out.size.x * grads out.size.y * grads out.size.z;
    grads out.size.y = 1;
    grads out.size.z = 1;
    for ( int b = 0; b < out.size.b; b++ ) {
        for ( int n = 0; n < activator input.size.x; n++ ){</pre>
            double ad = activator derivative( activator input(n, 0, 0, b) );
            double ng = grad next layer(n, 0, 0, b);
            act grad(n, 0, 0, b) = ad * ng;
    // Reorder loops and tile on n
    for ( int nn = 0; nn < out.size.x; nn+=BLOCK SIZE ) {</pre>
        for ( int b = 0; b < out.size.b; b++ ) {
            for ( int n = nn; n < nn + BLOCK SIZE && n < out.size.x; n++ ) {</pre>
                for ( int i = 0; i < grads out.size.x; i++ ) {
                    grads out(i, 0, 0, b) += act grad(n, 0, 0, b) * weights(i, n, 0);
    grads out.size = in.size;
```

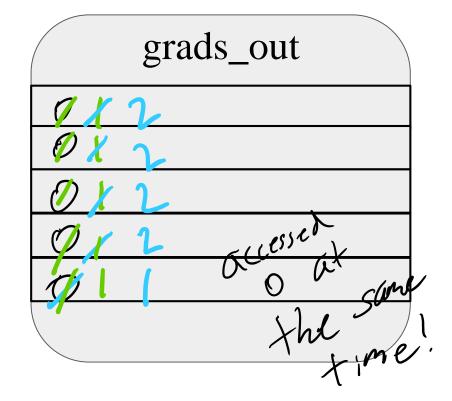
Multiple threads on nn loop

Multiple threads writing to the same memory location of grads_out

Multiple threads on nn loop issue

thread 0 🥕

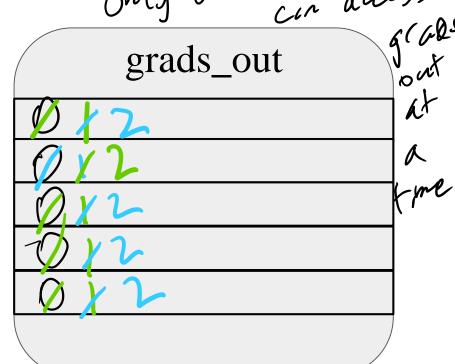
thread 1



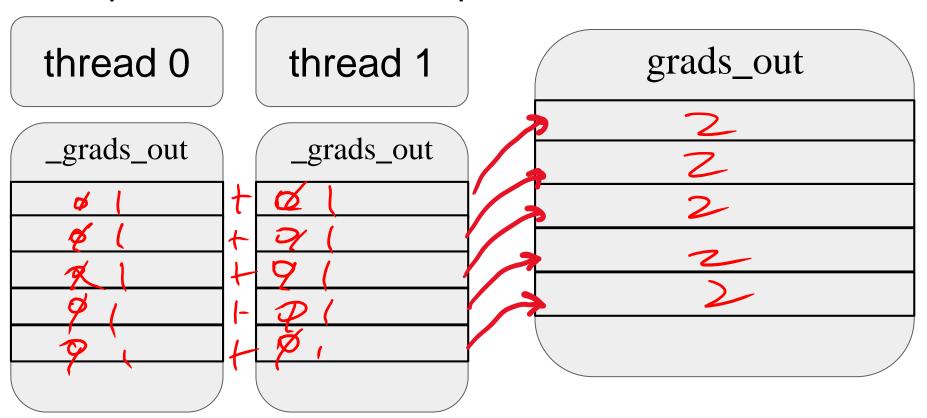
Multiple threads on nn loop simple fix on the auess can auess

thread 0

thread 1



Multiple threads on nn loop better fix



Multiple threads on b loop

- Each thread is writing to different locations of grads_out
 - More specifically, each thread handles writing to a different batch output
- No race conditions!

Multiple threads on n loop

- Is there a problem here?
- How can we fix it?
- Create local version of grads_out at the beginning of the loop and then sum results at the end of the loop!

Multiple threads on i loop

- Is there a problem here?
- No! We don't have to fix anything here. Each thread writes to distinct memory locations in grads_out.

Baseline - fc_fix_weights (Tier 2)

```
void fix weights() {
    tdsize old in size = in.size;
    in.size.x = in.size.x * in.size.y * in.size.z;
    in.size.y = 1;
    in.size.z = 1;
    for ( int b = 0; b < out.size.b; b++ ) {
        for ( int n = 0; n < weights.size.y; n++ ) {</pre>
            for ( int i = 0; i < weights.size.x; i++ ) {</pre>
                double& w = weights( i, n, 0 );
                double m = (act grad(n, 0, 0, b) + old act grad(n, 0, 0, b) * MOMENTUM);
                double g weight = w - (LEARNING RATE * m * in(i, 0, 0, b) + LEARNING RATE * WEIGHT DECAY * w);
                w = g weight;
            old act grad(n, 0, 0, b) = act grad(n, 0, 0, b) + old act grad(n, 0, 0, b) * MOMENTUM;
    in.size = old in size;
```

Multiple threads on b loop

```
for ( int b = 0; b < out.size.b; b++ ) {
    for ( int n = 0; n < weights.size.y; n++ ) {
        for ( int i = 0; i < weights.size.x; i++ ) {
            double& w = weights( i, n, 0 );
            double m = (act_grad(n, 0, 0, b) + old_act_grad(n, 0, 0, b) * MOMENTUM);
            double g_weight = w - (LEARNING_RATE * m * in(i, 0, 0, b) + LEARNING_RATE * WEIGHT_DECAY * w);
            w = g_weight;
        }
        old_act_grad(n, 0, 0, b) = act_grad(n, 0, 0, b) + old_act_grad(n, 0, 0, b) * MOMENTUM;
    }
}</pre>
```

- Is there a problem here?
- How can we fix it?
- Use omp critical around the accumulation.

Multiple threads on n loop

```
for ( int b = 0; b < out.size.b; b++ ) {
    for ( int n = 0; n < weights.size.y; n++ ) {
        for ( int i = 0; i < weights.size.x; i++ ) {
            double& w = weights( i, n, 0 );
            double m = (act_grad(n, 0, 0, b) + old_act_grad(n, 0, 0, b) * MOMENTUM);
            double g_weight = w - (LEARNING_RATE * m * in(i, 0, 0, b) + LEARNING_RATE * WEIGHT_DECAY * w);
            w = g_weight;
            }
            old_act_grad(n, 0, 0, b) = act_grad(n, 0, 0, b) + old_act_grad(n, 0, 0, b) * MOMENTUM;
        }
}</pre>
```

- Is there a problem here?
- No! We don't have to fix anything here. Each thread writes to distinct memory locations in weights and old_act_grad.

Multiple threads on i loop

```
for ( int b = 0; b < out.size.b; b++ ) {
    for ( int n = 0; n < weights.size.y; n++ ) {
        for ( int i = 0; i < weights.size.x; i++ ) {
            double& w = weights( i, n, 0 );
            double m = (act_grad(n, 0, 0, b) + old_act_grad(n, 0, 0, b) * MOMENTUM);
            double g_weight = w - (LEARNING_RATE * m * in(i, 0, 0, b) + LEARNING_RATE * WEIGHT_DECAY * w);
            w = g_weight;
            }
            old_act_grad(n, 0, 0, b) = act_grad(n, 0, 0, b) + old_act_grad(n, 0, 0, b) * MOMENTUM;
        }
}</pre>
```

- Is there a problem here?
- No! We don't have to fix anything here. Each thread writes to distinct memory locations in weights.

Tier 3 Hints

- Gprof does not work with multithreading
 - Comment out OMP=yes in config
 - Can't iteratively check gprof
- Don't be alarmed if you get slowdowns for some loops
- You'll need to do more than just adding multithreading to achieve the speedup of 6x
 - Try combining multithreading with previous optimizations like tiling and loop reordering
 - This is demonstrated in Tier 1

Questions?