#### Optimizations Assignment 2

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# **Bad Spatial Locality**

```
number of channels: 32..2048 (always powers of 2)
kernel order: 1, 3, 5, or 7
for (c = 0; c < nchannels; c++)
  for ( x = 0; x < kernel_order; x++) {
    for (y = 0; y < kernel_order; y++) {
      sum += image[w+x][h+y][c] * kernels[m][c][x][y];
  output[m][w][h] = (float) sum;
```

# **Better Spatial Locality**

```
number of channels: 32..2048 (always powers of 2)
kernel order: 1, 3, 5, or 7

for ( x = 0; x < kernel_order; x++) {
   for ( y = 0; y < kernel_order; y++ ) {
     for ( c = 0; c < nchannels; c++ ) {
        sum += image[w+x][h+y][c] * kernels[m][c][x][y];
     }
   }
   output[m][w][h] = (float) sum;
}</pre>
```

#### First Look At Loop unrolling

We can loop unroll our inner loop since nchannels can be big

```
for (x = 0; x < kernel_order; x++) {
  for (y = 0; y < kernel_order; y++) {
    for (c = 0; c < nchannels; c+=4)  { // power of 2
      sum + = image[w+x][h+y][c] * kernels[m][c][x][y];
      sum + = image[w+x][h+y][c+1]*kernels[m][c+1][x][y];
      sum + = image[w+x][h+y][c+2]*kernels[m][c+2][x][y];
      sum + = image[w+x][h+y][c+3]*kernels[m][c+3][x][y];
  output[m][w][h] = (float) sum;
```

# More Loop unrolling

```
for (x = 0; x < kernel_order; x++)
  for (y = 0; y < kernel_order; y++) {
    for ( c = 0; c < nchannels; c+=8) {
     sum + = image[w+x][h+y][c] * kernels[m][c][x][y];
      sum + = image[w+x][h+y][c+1]*kernels[m][c+1][x][y];
      sum + = image[w+x][h+y][c+2]*kernels[m][c+2][x][y];
     sum + = image[w+x][h+y][c+3]*kernels[m][c+3][x][y];
     sum + = image[w+x][h+y][c+4]*kernels[m][c+4][x][y];
      sum + = image[w+x][h+y][c+5]*kernels[m][c+5][x][y];
      sum + = image[w+x][h+y][c+6]*kernels[m][c+6][x][y];
     sum + = image[w+x][h+y][c+7]*kernels[m][c+7][x][y];
 output[m][w][h] = (float) sum;
```

# Even More Loop unrolling

```
for ( x = 0; x < kernel_order; x++) {
  for ( y = 0; y < kernel_order; y++ ) {
    for ( c = 0; c < nchannels; c+=16) {
      sum+=image[w+x][h+y][c]*kernels[m][c][x][y];
      sum+=image[w+x][h+y][c+1]*kernels[m][c+1][x][y];
      // 14 more times incrementing c index
    }
  }
  output[m][w][h] = (float)sum;
}</pre>
```

#### 1D Arrays

Every array will be return in 1D notation

```
for (x = 0; x < kernel_order; x++) {
  for (y = 0; y < kernel_order; y++) {
    for ( c = 0; c < nchannels; c+=16) {
      int image_offset = (w+x) * width_offset +
        ((h+y) * nchannels) + c;
      int kernel_total_offset =
       m * kernel_offset +
        x * kernel_order + y + c * ko2;
     sum += image_1d[image_offset] *
        kernel[kernel_total_offset];
     sum += image_1d[image_offset + 1] *
        kernel[kernel_total_offset + ko2 * 1];
```

#### Kernel Not In Contigous Memory

kernel isn't accessed in contiguous memory
sum += image\_1d[image\_offset] \*
 kernel[kernel\_total\_offset];
sum += image\_1d[image\_offset + 1] \*
 kernel[kernel\_total\_offset + ko2 \* 1];
sum += image\_1d[image\_offset + 2] \*
 kernel[kernel\_total\_offset + ko2 \* 2];

kernel[kernel\_total\_offset + ko2 \* 3];

sum += image\_1d[image\_offset + 3] \*

#### Transpose

Transposing kernel makes access times faster for inner loop

```
for (int m = 0; m < nkernels; m++) {
  for (int c = 0; c < nchannels; c++) {
    for (int x = 0; x < kernel_order; x++) {
      m_times_kernel_offset = m * kernel_offset;
      c_{times_ko2} = c * ko2;
      x_times_kernel_order = x * kernel_order;
      kernel_total_offset_precalc =
        m times kernel offset +
        x_times_kernel_order + c_times_ko2;
      t_kernel_total_offset_precalc =
        m_times_kernel_offset +
        x_times_kernel_order * nchannels + c:
```

#### Loop Fusing/Loop Collapsing

Transposing kernel makes access times faster for inner loop

```
for(int n = 0; n < (nkernels*nchannels*kernel_order);</pre>
  int m = n/(nchannels*kernel_order);
  int c = (n%(nchannels*kernel_order))/kernel_order;
  int x = (n%(nchannels*kernel_order))%kernel_order;
  m_times_kernel_offset = m * kernel_offset;
  c_{times_ko2} = c * ko2;
  x_times_kernel_order = x * kernel_order;
  kernel_total_offset_precalc =
    m times kernel offset +
    x_times_kernel_order + c_times_ko2;
  t_kernel_total_offset_precalc =
    m_times_kernel_offset +
    x_times_kernel_order * nchannels + c;
```

# Continued: Loop Fusing/Loop Collapsing

```
for(int y = 0; y < kernel_order; y++) {
    t_kernel[t_kernel_total_offset_precalc] =
        (double) kernel[kernel_total_offset_precalc+y];
    t_kernel_total_offset_precalc += nchannels;
}</pre>
```

# Parallel Transposing

#### Parallelise using omp

```
#pragma omp parallel for
for(int n = 0; n < (nkernels*nchannels*kernel_order);
int m = n/(nchannels*kernel_order);
int c = (n%(nchannels*kernel_order))/kernel_order;
int x = (n%(nchannels*kernel_order))%kernel_order;
...
}</pre>
```

# Parallel and Collapse Main Convolution

```
Parallelise using omp
#pragma omp parallel for
for (int n = 0; n < nkernels*width*height; n++) {
  int m = n/(width*height);
  int w = (n%(width*height))/height;
  int h = (n%(width*height))%height;
  ...
}</pre>
```

#### Align Kernel array

Aligning memory ensures better cache hits. This will also allow us to vectors in the next slide.

```
double * t_kernel =
   _mm_malloc(sizeof(double) *
    nchannels * kernel_order *
    kernel_order * nkernels, 16); // 16 byte aligned
```

#### Vectorisation

Unfortunately floats cause issues when summed up as the result may overflow, hence we are forced to work with

```
_{-m}128d v4sum = _{mm_setzero_pd();}
// 2 kernel_order for loops...
for (c = 0; c < nchannels; c+=2)
  _{-m}128 \text{ v4image}_{-1}d =
    _mm_loadu_ps(image_1d+image_offset+c);
  _{-m}128d v4image_{-}1d_{-}pd =
    _mm_cvtps_pd(v4image_1d);
  _m128d v4t_kernel_pd =
    _mm_load_pd(t_kernel+kernel_total_offset+c);
  _{-m}128d product =
    _mm_mul_pd(v4image_1d_pd, v4t_kernel_pd);
  v4sum = _mm_add_pd(v4sum, product);
```

#### Continuation: Vectorisation

```
c += 2:
v4image_1d =
  _mm_shuffle_ps(v4image_1d, v4image_1d,
  _MM_SHUFFLE(0, 0, 3, 2));
v4image_1d_pd = _mm_cvtps_pd(v4image_1d);
v4t_kernel_pd =
  _mm_load_pd(t_kernel+kernel_total_offset+c);
product = _mm_mul_pd(v4image_1d_pd, v4t_kernel_pd);
v4sum = _mm_add_pd(v4sum, product);
// repeat inner loop 12 more times
```

#### Final Step in Vectorisation

We must finally sum up, after finishing the kernel convolution We repeat for every nkernels again.

Sum up using hadd\_pd, which adds two double vectors together Extract lower double using cvtsd function

```
v4sum = _mm_hadd_pd(v4sum, v4sum);
output1d[m * width_times_height + w * height + h] =
  (float) _mm_cvtsd_f64(v4sum);
```

#### Source Code

```
void student_conv(float *** restrict image, int16_t **** restrict kernels, float *** restrict output,
       int width, int height, int nchannels, int nkernels.
       int kernel order)
   int width times height = width * height;
   float * output1d = **output;
   float * image 1d = **image:
   int16_t * kernel = ***kernels;
   int ko2 = kernel order * kernel order:
   int width offset = (height+kernel order) * nchannels;
   int kernel_offset = nchannels * ko2;
   int image_offset_precalc;
   int kernel total offset precalc:
   int image offset;
   int kernel total offset:
   double * t_kernel = _mm_malloc(sizeof(double) * nchannels * kernel_order * kernel_order * nkernels, 16);
   int m times kernel offset;
   int c_times_ko2;
   int x times kernel order:
   int t_kernel_total_offset_precalc;
   for(int n = 0: n < (nkernels*nchannels*kernel order): n++){</pre>
       int m = n/(nchannels*kernel order);
       int c = (n%(nchannels*kernel order))/kernel order:
       int x = (n%(nchannels*kernel order))%kernel order;
       m times kernel_offset = m * kernel_offset;
       c times ko2 = c * ko2:
       x times kernel order = x * kernel order;
       kernel total offset precalc = m times kernel offset + x times kernel order + c times ko2:
       t kernel total offset precalc = m times kernel offset + x times kernel order * nchannels + c;
       for(int y = 0; y < kernel_order; y++){
           t kernel[t kernel total offset precalc] = (double) kernel[kernel total offset precalc+v]:
           t kernel total offset precalc += nchannels;
```

#### Continued: Source Code

```
m_times_kernel_offset = m * kernel_offset;
__m128d v4sum = _mm_setzero pd()
    image_offset_precalc = (w+x) * width_offset;
   kernel total offset precalc = m times kernel offset + x * kernel order * nchannels;
    for ( y = 0; y < kernel_order; y++ )
        image offset = image offset precalc + (h+v) * nchannels:
        kernel_total_offset = kernel_total_offset_precalc + y * nchannels;
            __m128 v4image_ld = _mm_loadu_ps(image_ld+image_offset+c);
            m128d v4image_ld_pd = _mm_cvtps_pd(v4image_ld);
            m128d v4t kernel pd = mm load pd(t kernel+kernel total offset+c);
            __m128d product = _mm_mul_pd(v4image_ld_pd, v4t_kernel_pd);
            v4sum = _mm_add_pd(v4sum, product);
            v4image_1d = _mm_shuffle_ps(v4image_1d, v4image_1d, _MM_SHUFFLE(0, 0, 3, 2));
            v4image 1d pd - mm cvtps pd(v4image 1d);
            v4t kernel pd = mm load pd(t kernel+kernel total offset+c):
           product = _mm_mul_pd(v4image_ld_pd, v4t_kernel_pd);
v4sum = _mm_add_pd(v4sum, product);
            v4image_ld = _mm_loadu_ps(image_ld+image_offset+c);
            v4image_ld_pd = _mm_cvtps_pd(v4image_ld);
            product = _mm_mul_pd(v4image_1d_pd, v4t_kernel_pd);
            v4image_ld = _mm_shuffle_ps(v4image_ld, v4image_ld, _MM_SHUFFLE(0, 0, 3, 2));
            v4image_ld_pd = _mm_cvtps_pd(v4image_ld);
            product = _mm_mul_pd(v4image_ld_pd, v4t_kernel_pd);
            v4sum = _mm_add_pd(v4sum, product);
            v4image_id = _mm_loadu_ps(image_id+image_offset+c);
            v4image 1d pd = mm cvtps pd(v4image 1d):
            v4t_kernel_pd = _mm_load_pd(t_kernel+kernel_total_offset+c);
            product = _mm_mul_pd(v4image_ld_pd, v4t_kernel_pd);
            v4sum = mm add od(v4sum, product):
            v4image 1d = mm shuffle ps(v4image 1d, v4image 1d, MM SHUFFLE(0, 0, 3, 2))
```

#### Continued: Source Code

```
v4image 1d pd = _mm_cvtps_pd(v4image_1d);
            v4t kernel pd = mm load pd(t kernel+kernel total offset+c);
            product = _mm_mul_pd(v4image_1d_pd, v4t_kernel_pd);
            v4sum = mm add pd(v4sum, product);
            v4image 1d = mm loadu ps(image 1d+image offset+c):
            v4image 1d pd = mm cvtps pd(v4image 1d);
            v4t kernel pd = mm load pd(t kernel+kernel total offset+c):
            product = mm mul pd(v4image 1d pd, v4t kernel pd);
            v4sum = _mm_add_pd(v4sum, product);
            v4image_1d = _mm_shuffle_ps(v4image_1d, v4image_1d, _MM_SHUFFLE(0, 0, 3, 2));
            v4image_1d_pd = _mm_cvtps_pd(v4image_1d);
            v4t kernel pd = mm load pd(t kernel+kernel total offset+c);
            product = _mm_mul_pd(v4image_1d_pd, v4t_kernel_pd);
            v4sum = _mm_add_pd(v4sum, product);
v4sum = _mm_hadd_pd(v4sum, v4sum); // add the two lanes together and put in lower lane
output1d[m * width times height + w * height + h] = (float) mm cvtsd f64(v4sum): // extract lower double
```

#### Speed Increase

With test conv 128 128 7 256 256, we can get to nearly 90x speed increase.

sepelena@stoker:~/Concurrency\$ make run

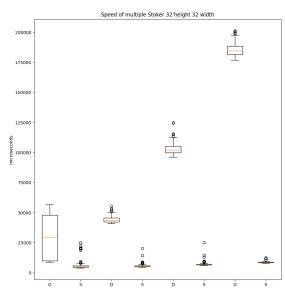
Executing: conv

David conv time: 134442611 microseconds
Student conv time: 1497555 microseconds

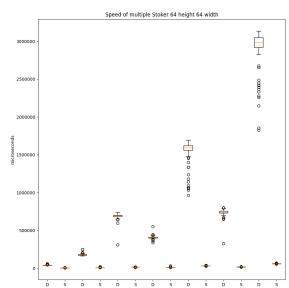
The total speed up time was 89.77x and 132945056 microseconds less

COMMENT: sum of absolute differences (0.000000) within acceptable range (0.062500)

# Graphing up to 128 kernel\_orders



# Graphing up to 128 kernel\_orders



# Graphing up to 128 kernel\_orders

