





THE UNIVERSITY of EDINBURGH **informatics**

- Largest Informatics Department in the UK:
 - > 500 academic and research staff + PhD students
- Overall 6 Research Institutes
 - 2 particular relevant for the topic of the talk:
- ICSA — Institute for Computing Systems Architecture
 - Compiler & Architecture
 - Parallel Computing
 - ...
- LFCS — Laboratory for Foundations of Computer Science
 - Programming Languages and Foundations
 - Software Engineering
 - ...



Structured Parallel Programming

From High-Level Functional Expressions

to High-Performance OpenCL Code

Michel Steuwer

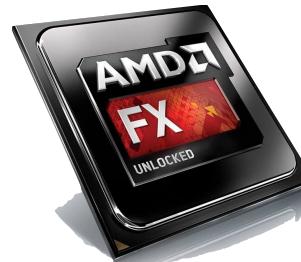
<http://homepages.inf.ed.ac.uk/msteuwer/>



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The Problem(s)

- Parallel processors everywhere



CPU

- Many different types: CPUs, GPUs, ...



GPU

- **First Major Challenge:**
Parallel programming is hard.
Optimising is even harder!



FPGA

- **Second Major Challenges:**
No portability of performance!

Accelerator



Part I: Addressing the Programmability Challenge



Programming with OpenCL

- Case Study: Parallel reduction in OpenCL

```
kernel void reduce(global float* g_idata, global float* g_odata,
                  unsigned int n, local float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i   = get_global_id(0);
    l_data[tid] = (i < n) ? g_idata[i] : 0;
    barrier(CLK_LOCAL_MEM_FENCE);
    // do reduction in local memory
    for (unsigned int s=1; s < get_local_size(0); s*= 2) {
        if ((tid % (2*s)) == 0) {
            l_data[tid] += l_data[tid + s];
            barrier(CLK_LOCAL_MEM_FENCE);
        }
    }
    // write result for this work-group to global memory
    if (tid == 0) g_odata[get_group_id(0)] = l_data[0];
}
```



Programming with OpenCL

- Case Study: Parallel reduction in OpenCL

Kernel function executed in parallel by multiple **work-items**

```
kernel void reduce(global float* g_idata, global float* g_odata,
                  unsigned int n, local float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i   = get_global_id(0);
    l_data[tid] = (i < n) ? g_idata[i] : 0;
    barrier(CLK_LOCAL_MEM_FENCE);
    // do reduction in local memory
    for (unsigned int s=1; s < get_local_size(0); s*= 2) {
        if ((tid % (2*s)) == 0) {
            l_data[tid] += l_data[tid + s];
            barrier(CLK_LOCAL_MEM_FENCE);
        }
    }
    // write result for this work-group to global memory
    if (tid == 0) g_odata[get_group_id(0)] = l_data[0];
}
```

Work-items are identified by a unique **global id**



Programming with OpenCL

- Case Study: Parallel reduction in OpenCL

Work-items are grouped into work-groups

Local id within work-group

```
kernel void reduce(global float* g_idata, global float* g_odata,
                  unsigned int n, local float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i   = get_global_id(0);
    l_data[tid] = (i < n) ? g_idata[i] : 0;
    barrier(CLK_LOCAL_MEM_FENCE);
    // do reduction in local memory
    for (unsigned int s=1; s < get_local_size(0); s*= 2) {
        if ((tid % (2*s)) == 0) {
            l_data[tid] += l_data[tid + s];
            barrier(CLK_LOCAL_MEM_FENCE);
        }
    }
    // write result for this work-group to global memory
    if (tid == 0) g_odata[get_group_id(0)] = l_data[0];
}
```



Programming with OpenCL

- Case Study: Parallel reduction in OpenCL

Big, but slow **global** memory

Small, but fast **local** memory

```
kernel void reduce(global float* g_idata, global float* g_odata,
                  unsigned int n, local float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i   = get_global_id(0);
    l_data[tid] = (i < n) ? g_idata[i] : 0;
    barrier(CLK_LOCAL_MEM_FENCE);
    // do reduction in local memory
    for (unsigned int s=1; s < get_local_size(0); s*= 2) {
        if ((tid % (2*s)) == 0) {
            l_data[tid] += l_data[tid + s];
            barrier(CLK_LOCAL_MEM_FENCE);
        }
    }
    // write result for this work-group to global memory
    if (tid == 0) g_odata[get_group_id(0)] = l_data[0];
}
```

Memory **barriers** for consistency



Programming with OpenCL

- Case Study: Parallel reduction in OpenCL

```
kernel void reduce(global float* g_idata, global float* g_odata,
                  unsigned int n, local float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i   = get_global_id(0);
    l_data[tid] = (i < n) ? g_idata[i] : 0;
    barrier(CLK_LOCAL_MEM_FENCE);
    // do reduction in local memory
    for (unsigned int s=1; s < get_local_size(0); s*= 2) {
        if ((tid % (2*s)) == 0) {
            l_data[tid] += l_data[tid + s];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    // write result for this work-group to global memory
    if (tid == 0) g_odata[get_group_id(0)] = l_data[0];
}
```

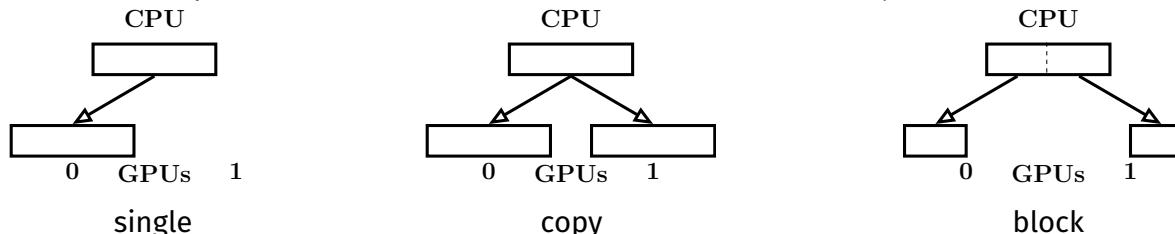
Functionally correct implementations in OpenCL are hard!



The SkelCL Programming Model

Three high-level features added to OpenCL:

- **parallel container data types**
for unified memory management between CPU and (multiple) GPUs
 - **implicit memory transfers between CPU and GPU**
 - **lazy copying minimizes data transfers**
- **recurring patterns of parallelism**
(a.k.a., algorithmic skeletons) for easily expressing parallel computation patterns;
 $\text{zip } (\oplus) [x_1, \dots, x_n] [y_1, \dots, y_n] = [x_1 \oplus y_1, \dots, x_n \oplus y_n]$
 $\text{reduce } (\oplus) \oplus_{\text{id}} [x_1, \dots, x_n] = \oplus_{\text{id}} \oplus x_1 \oplus \dots \oplus x_n$
- **data distribution and redistribution**
mechanisms for transparent data transfers in multi-GPU systems.



The SkelCL Library by Example

$\text{dotProduct } A \text{ } B = \text{reduce } (+) \text{ } 0 \circ \text{zip } (\times) \text{ } A \text{ } B$

```
#include <SkelCL/SkelCL.h>
#include <SkelCL/Zip.h>
#include <SkelCL/Reduce.h>
#include <SkelCL/Vector.h>

float dotProduct(const float* a, const float* b, int n) {
    using namespace skelcl;
    skelcl::init( 1_device.type(deviceType::ANY) );

    auto mult =    zip([](float x, float y) { return x*y; });
    auto sum  = reduce([](float x, float y) { return x+y; }, 0);

    Vector<float> A(a, a+n); Vector<float> B(b, b+n);

    Vector<float> C = sum( mult(A, B) );

    return C.front();
}
```



From SkelCL to OpenCL

1

```
#include <SkelCL/SkelCL.h>
#include <SkelCL/Zip.h>
#include <SkelCL/Reduce.h>
#include <SkelCL/Vector.h>

float dotProduct(const float* a, const float* b, int n) {
    using namespace skelcl;
    skelcl::init( 1_device.type(deviceType::ANY) );

    auto mult =    zip([](float x, float y) { return x*y; });
    auto sum   = reduce([](float x, float y) { return x+y; }, 0);

    Vector<float> A(a, a+n); Vector<float> B(b, b+n);

    Vector<float> C = sum( mult(A, B) );

    return C.front();
}
```

1

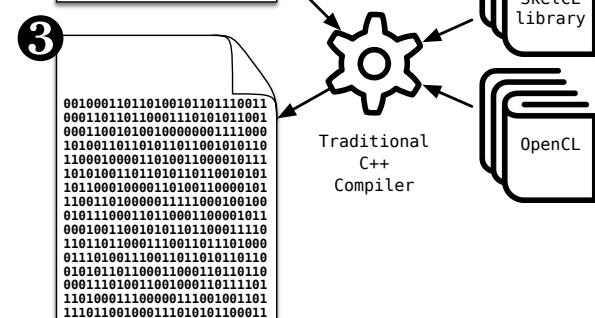
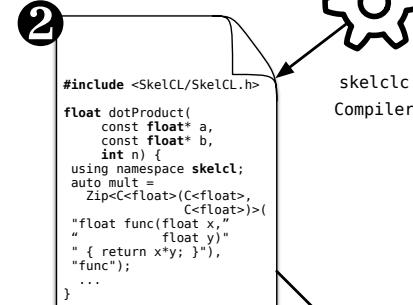
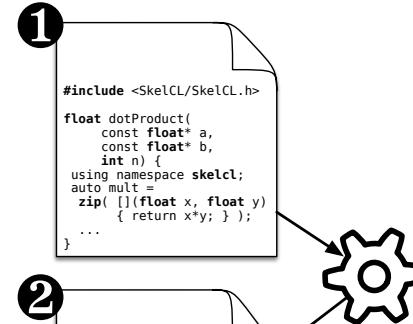
```
#include <SkelCL/SkelCL.h>
float dotProduct(
    const float* a,
    const float* b,
    int n) {
    using namespace skelcl;
    auto mult =
        zip( [] (float x, float y)
            { return x*y; } );
    ...
}
```

2

```
#include <SkelCL/SkelCL.h>
float dotProduct(
    const float* a,
    const float* b,
    int n) {
    using namespace skelcl;
    auto mult =
        Zip<Cfloat>(Cfloat>,
                     Cfloat>)>(
    "float func(float x,"
     "           float y)"
    " { return x*y; }",
    "func");
    ...
}
```

3

```
001000110110100101101110011
000110110110001110101011001
000110010100100000001110000
1010001101101010101001010110
10100000001101001100000000000
1010100110110101010110010101
10100000000000000000000000000
10110001000000000000000000000
11001100000000000000000000000
01011100011010000110000000000
00010011000101010101000000000
01011100011010000110000000000
11011011000000000000000000000
011101001110001101010101010
01010110110000110000110101010
0001101001000110000110111101
11010001110000001100010000000
11010011000000000000000000000
```



From SkelCL to OpenCL

2

```
#include <SkelCL/SkelCL.h>
#include <SkelCL/Zip.h>
#include <SkelCL/Reduce.h>
#include <SkelCL/Vector.h>

float dotProduct(const float* a, const float* b, int n) {
    using namespace skelcl;
    skelcl::init( 1_device.type(deviceType::ANY) );

    auto mult = Zip<Container<float>>(Container<float>,
                                         Container<float>)(>(
        Source("float func(float x, float y) {return x*y;};"));
    auto sum = Reduce<Vector<float>>(Vector<float>)(>(
        Source("float func(float x, float y) {return x+y;};"), "0");

    Vector<float> A(a, a+n); Vector<float> B(b, b+n);

    Vector<float> C = sum( mult(A, B) );

    return C.front();
}
```

1

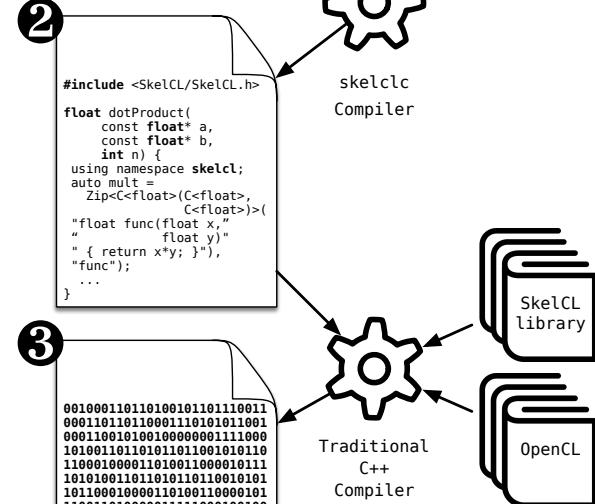
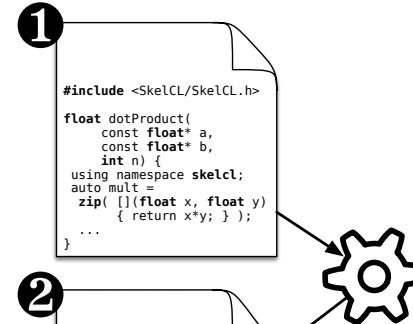
```
#include <SkelCL/SkelCL.h>
float dotProduct(
    const float* a,
    const float* b,
    int n) {
    using namespace skelcl;
    auto mult =
        zip( [] (float x, float y)
            { return x*y; } );
    ...
}
```

2

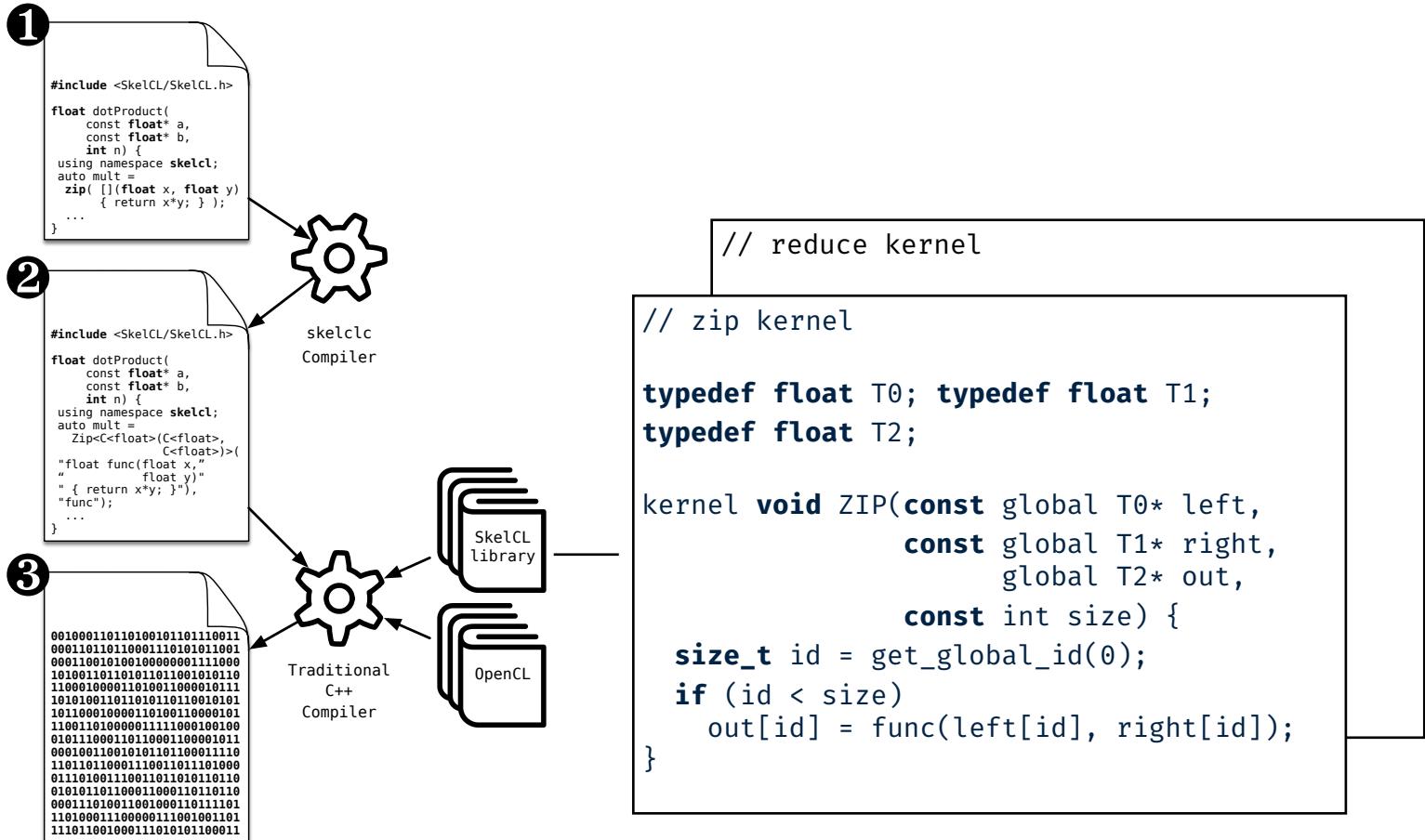
```
#include <SkelCL/SkelCL.h>
float dotProduct(
    const float* a,
    const float* b,
    int n) {
    using namespace skelcl;
    auto mult =
        Zip<C<float>,>(>(
            "float func(float x,"
            "           float y)"
            "           { return x*y; };",
            "func");
    ...
}
```

3

```
001000110110100101101110011
00011011011000111010101001
00011001010010000000111000
1010011011010101010100101010
1010010001101001010000000000
1100000000110100101000000000
1010100110000101010100000000
1011000100000101010100000000
1100110100000111000000000000
0001001100010101010100000000
0101110001101000011000000000
0001001100011000000000000000
01110100111000110010101010
0101011011000110000000000000
0001101001000110000000000000
110100011000000111001001001
1101001000110000000000000000
```



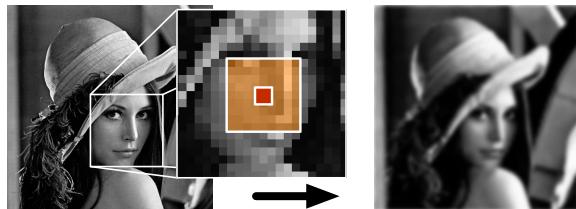
From SkelCL to OpenCL



Two Novel Algorithmic Skeletons

Stencil Computations

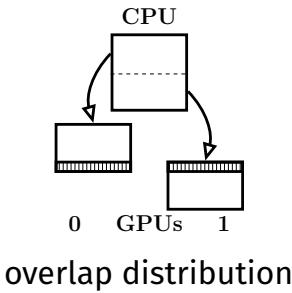
Example: Gaussian blur



$\text{gauss } M = \text{stencil } f \ 1 \ \bar{0} \ M$

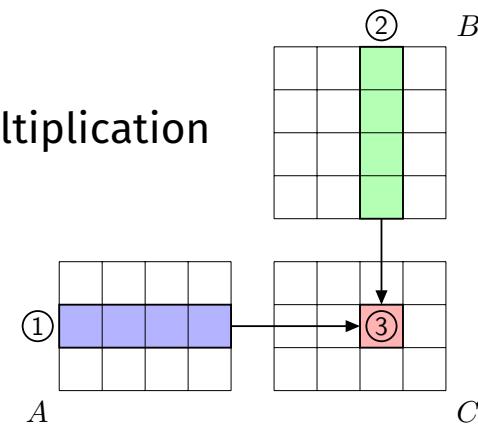
where f is the weighted gaussian kernel

Multi-GPU support:



Allpairs Computations

Example:
Matrix Multiplication



$A \times B = \text{allpairs dotProduct } A \ B^T$

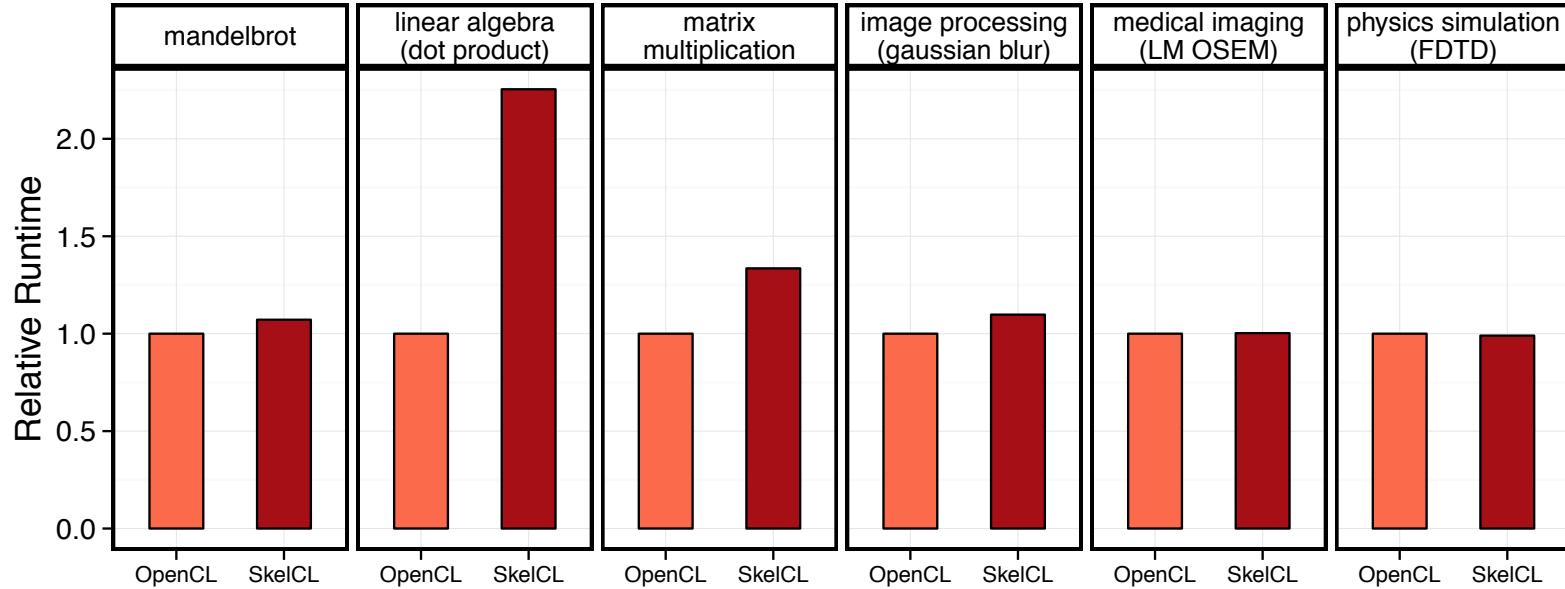
Optimization for zipReduce patterns:

$\text{dotProduct } a \ b = \text{zipReduce } (+) \ 0 \ (x) \ a \ b$

Multi-GPU support with
block and **copy** distribution



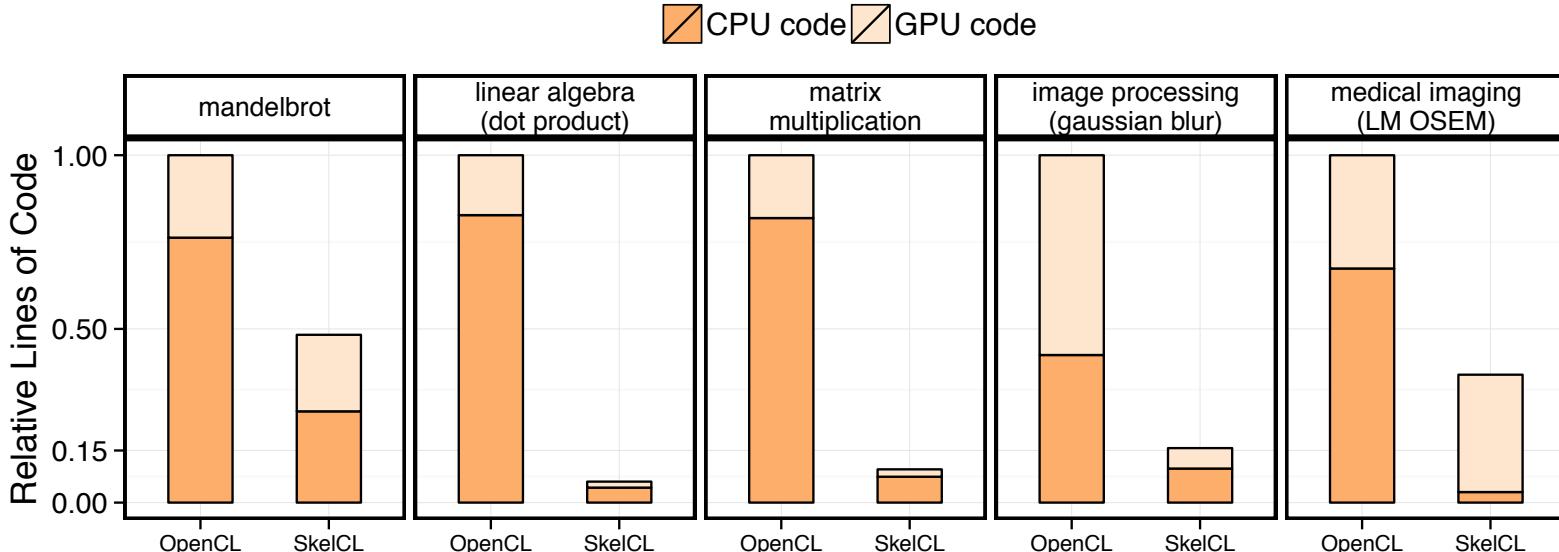
SkelCL Evaluation – Performance



SkelCL performance close to native OpenCL code!

(Exception: dot product ... wait for Part II)

SkelCL Evaluation – Productivity



SkelCL programs are significantly shorter!

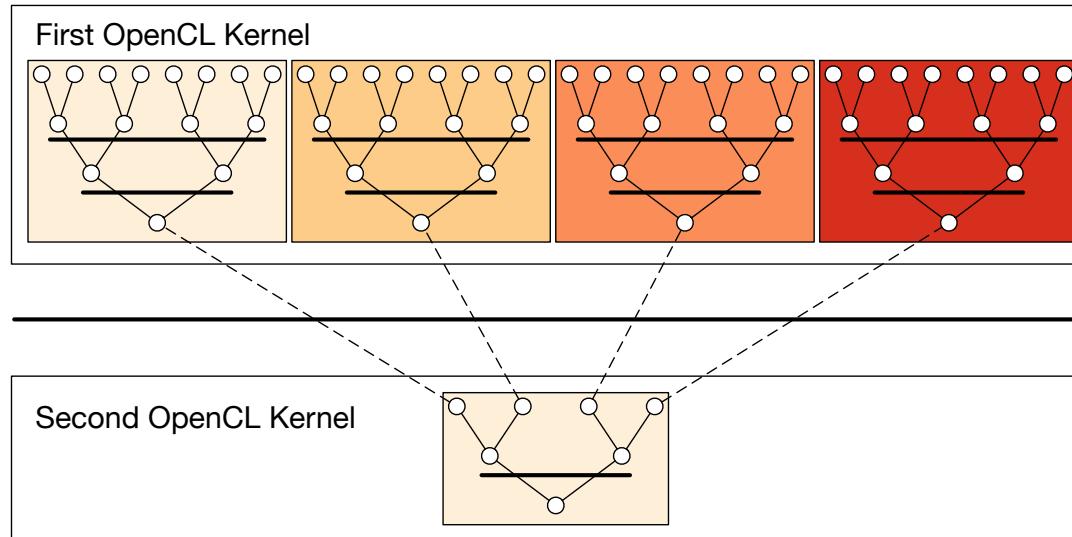
SkelCL is open source software and available from
<http://github.com/skelcl/skelcl>

Part II: Addressing the Performance Portability Challenge



Case Study: Parallel Reduction in OpenCL

- Summing up all values of an array
- Comparison of 7 implementations by Nvidia
- Investigating complexity and efficiency of optimisations



Unoptimised Implementation Parallel Reduction

```
kernel void reduce0(global float* g_idata, global float* g_odata,
                    unsigned int n, local float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i   = get_global_id(0);
    l_data[tid] = (i < n) ? g_idata[i] : 0;
    barrier(CLK_LOCAL_MEM_FENCE);
    // do reduction in local memory
    for (unsigned int s=1; s < get_local_size(0); s*= 2) {
        if ((tid % (2*s)) == 0) {
            l_data[tid] += l_data[tid + s];
            barrier(CLK_LOCAL_MEM_FENCE);
        }
    }
    // write result for this work-group to global memory
    if (tid == 0) g_odata[get_group_id(0)] = l_data[0];
}
```



Avoid Divergent Branching

```
kernel void reduce1(global float* g_idata, global float* g_odata,
                    unsigned int n, local float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i   = get_global_id(0);
    l_data[tid] = (i < n) ? g_idata[i] : 0;
    barrier(CLK_LOCAL_MEM_FENCE);

    for (unsigned int s=1; s < get_local_size(0); s*= 2) {
        // continuous work-items remain active
        int index = 2 * s * tid;
        if (index < get_local_size(0)) {
            l_data[index] += l_data[index + s];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (tid == 0) g_odata[get_group_id(0)] = l_data[0];
}
```



Avoid Interleaved Addressing

```
kernel void reduce2(global float* g_idata, global float* g_odata,
                    unsigned int n, local float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i   = get_global_id(0);
    l_data[tid] = (i < n) ? g_idata[i] : 0;
    barrier(CLK_LOCAL_MEM_FENCE);

    // process elements in different order
    // requires commutativity
    for (unsigned int s=get_local_size(0)/2; s>0; s>>=1) {
        if (tid < s) {
            l_data[tid] += l_data[tid + s];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (tid == 0) g_odata[get_group_id(0)] = l_data[0];
}
```



Increase Computational Intensity per Work-Item

```
kernel void reduce3(global float* g_idata, global float* g_odata,
                    unsigned int n, local float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i = get_group_id(0) * (get_local_size(0)*2)
                    + get_local_id(0);
    l_data[tid] = (i < n) ? g_idata[i] : 0;
    // performs first addition during loading
    if (i + get_local_size(0) < n)
        l_data[tid] += g_idata[i+get_local_size(0)];
    barrier(CLK_LOCAL_MEM_FENCE);

    for (unsigned int s=get_local_size(0)/2; s>0; s>>=1) {
        if (tid < s) {
            l_data[tid] += l_data[tid + s];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (tid == 0) g_odata[get_group_id(0)] = l_data[0];
}
```



Avoid Synchronisation inside a Warp

```
kernel void reduce4(global float* g_idata, global float* g_odata,
                    unsigned int n, local volatile float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i = get_group_id(0) * (get_local_size(0)*2)
                    + get_local_id(0);
    l_data[tid] = (i < n) ? g_idata[i] : 0;
    if (i + get_local_size(0) < n)
        l_data[tid] += g_idata[i+get_local_size(0)];
    barrier(CLK_LOCAL_MEM_FENCE);

    # pragma unroll 1
    for (unsigned int s=get_local_size(0)/2; s>32; s>>=1) {
        if (tid < s) { l_data[tid] += l_data[tid + s]; }
        barrier(CLK_LOCAL_MEM_FENCE); }

    // this is not portable OpenCL code!
    if (tid < 32) {
        if (WG_SIZE >= 64) { l_data[tid] += l_data[tid+32]; }
        if (WG_SIZE >= 32) { l_data[tid] += l_data[tid+16]; }
        if (WG_SIZE >= 16) { l_data[tid] += l_data[tid+ 8]; }
        if (WG_SIZE >=  8) { l_data[tid] += l_data[tid+ 4]; }
        if (WG_SIZE >=  4) { l_data[tid] += l_data[tid+ 2]; }
        if (WG_SIZE >=  2) { l_data[tid] += l_data[tid+ 1]; } }

    if (tid == 0) g_odata[get_group_id(0)] = l_data[0]; }
```

Complete Loop Unrolling

```
kernel void reduce5(global float* g_idata, global float* g_odata,
                    unsigned int n, local volatile float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i = get_group_id(0) * (get_local_size(0)*2)
                    + get_local_id(0);
    l_data[tid] = (i < n) ? g_idata[i] : 0;
    if (i + get_local_size(0) < n)
        l_data[tid] += g_idata[i+get_local_size(0)];
    barrier(CLK_LOCAL_MEM_FENCE);

    if (WG_SIZE >= 256) {
        if (tid < 128) { l_data[tid] += l_data[tid+128]; }
        barrier(CLK_LOCAL_MEM_FENCE); }

    if (WG_SIZE >= 128) {
        if (tid < 64) { l_data[tid] += l_data[tid+ 64]; }
        barrier(CLK_LOCAL_MEM_FENCE); }

    if (tid < 32) {
        if (WG_SIZE >= 64) { l_data[tid] += l_data[tid+32]; }
        if (WG_SIZE >= 32) { l_data[tid] += l_data[tid+16]; }
        if (WG_SIZE >= 16) { l_data[tid] += l_data[tid+ 8]; }
        if (WG_SIZE >= 8) { l_data[tid] += l_data[tid+ 4]; }
        if (WG_SIZE >= 4) { l_data[tid] += l_data[tid+ 2]; }
        if (WG_SIZE >= 2) { l_data[tid] += l_data[tid+ 1]; } }

    if (tid == 0) g_odata[get_group_id(0)] = l_data[0]; }
```

Fully Optimised Implementation

```
kernel void reduce6(global float* g_idata, global float* g_odata,
                    unsigned int n, local volatile float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i = get_group_id(0) * (get_local_size(0)*2)
                    + get_local_id(0);
    unsigned int gridSize = WG_SIZE * get_num_groups(0);
    l_data[tid] = 0;
    while (i < n) { l_data[tid] += g_idata[i];
                      if (i + WG_SIZE < n)
                          l_data[tid] += g_idata[i+WG_SIZE];
                      i += gridSize; }
    barrier(CLK_LOCAL_MEM_FENCE);

    if (WG_SIZE >= 256) {
        if (tid < 128) { l_data[tid] += l_data[tid+128]; }
        barrier(CLK_LOCAL_MEM_FENCE); }

    if (WG_SIZE >= 128) {
        if (tid < 64) { l_data[tid] += l_data[tid+ 64]; }
        barrier(CLK_LOCAL_MEM_FENCE); }

    if (tid < 32) {
        if (WG_SIZE >= 64) { l_data[tid] += l_data[tid+32]; }
        if (WG_SIZE >= 32) { l_data[tid] += l_data[tid+16]; }
        if (WG_SIZE >= 16) { l_data[tid] += l_data[tid+ 8]; }
        if (WG_SIZE >=  8) { l_data[tid] += l_data[tid+ 4]; }
        if (WG_SIZE >=  4) { l_data[tid] += l_data[tid+ 2]; }
        if (WG_SIZE >=  2) { l_data[tid] += l_data[tid+ 1]; } }

    if (tid == 0) g_odata[get_group_id(0)] = l_data[0]; }
```

Case Study Conclusions

- Optimising OpenCL is complex
 - Understanding of target hardware required
- Program changes not obvious
- Is it worth it? ...

```
kernel
void reduce0(global float* g_idata,
             global float* g_odata,
             unsigned int n,
             local float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i   = get_global_id(0);
    l_data[tid] = (i < n) ? g_idata[i] : 0;
    barrier(CLK_LOCAL_MEM_FENCE);

    for (unsigned int s=1;
         s < get_local_size(0); s*= 2) {
        if ((tid % (2*s)) == 0) {
            l_data[tid] += l_data[tid + s];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (tid == 0)
        g_odata[get_group_id(0)] = l_data[0];
}
```

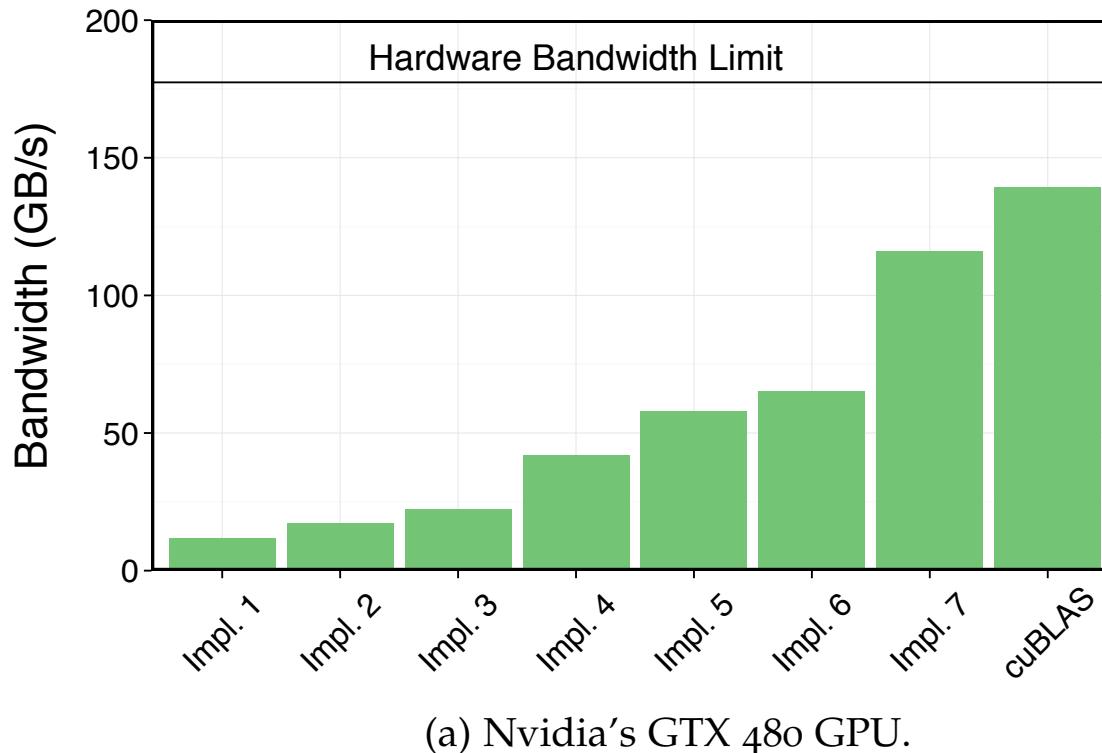
Unoptimized Implementation

```
kernel
void reduce6(global float* g_idata,
             global float* g_odata,
             unsigned int n,
             local volatile float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i =
        get_group_id(0) * (get_local_size(0)*2)
        + get_local_id(0);
    unsigned int gridSize =
        WG_SIZE * get_num_groups(0);
    l_data[tid] = 0;
    while (i < n) {
        l_data[tid] += g_idata[i];
        if (i + WG_SIZE < n)
            l_data[tid] += g_idata[i+WG_SIZE];
        i += gridSize;
    }
    barrier(CLK_LOCAL_MEM_FENCE);

    if (WG_SIZE >= 256) {
        if (tid < 128) {
            l_data[tid] += l_data[tid+128];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (WG_SIZE >= 128) {
        if (tid < 64) {
            l_data[tid] += l_data[tid+ 64];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (tid < 32) {
        if (WG_SIZE >= 64) {
            l_data[tid] += l_data[tid+32];
        }
        if (WG_SIZE >= 32) {
            l_data[tid] += l_data[tid+16];
        }
        if (WG_SIZE >= 16) {
            l_data[tid] += l_data[tid+ 8];
        }
        if (WG_SIZE >= 8) {
            l_data[tid] += l_data[tid+ 4];
        }
        if (WG_SIZE >= 4) {
            l_data[tid] += l_data[tid+ 2];
        }
        if (WG_SIZE >= 2) {
            l_data[tid] += l_data[tid+ 1];
        }
    }
    if (tid == 0)
        g_odata[get_group_id(0)] = l_data[0];
}
```

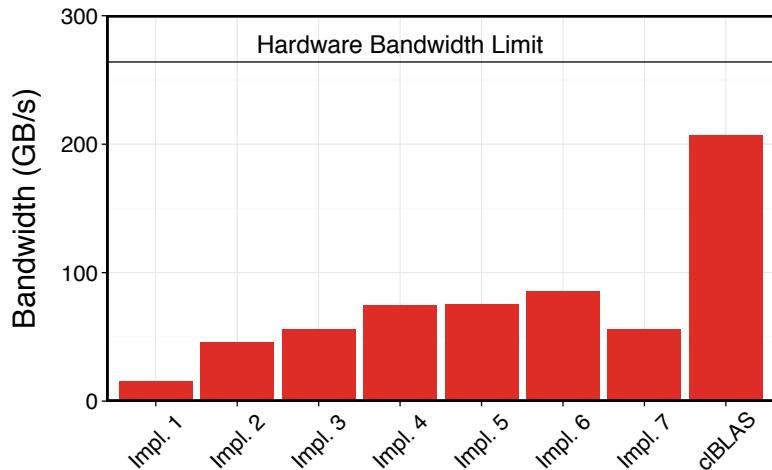
Fully Optimized Implementation

Performance Results Nvidia

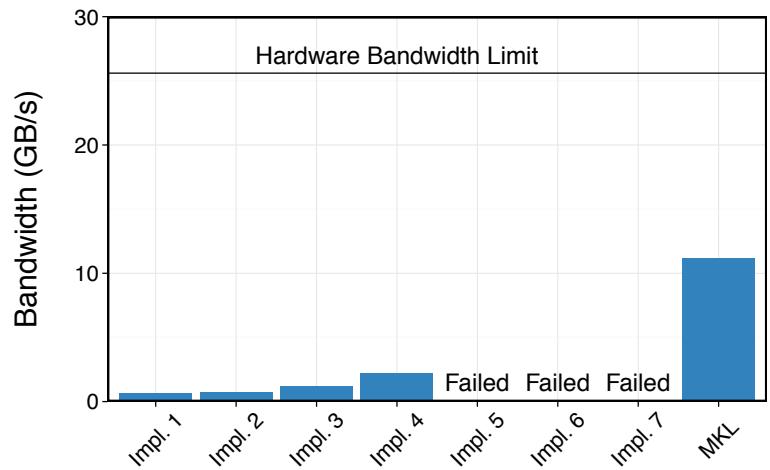


- ... Yes! Optimising improves performance by a factor of 10!
- Optimising is important, but ...

Performance Results AMD and Intel



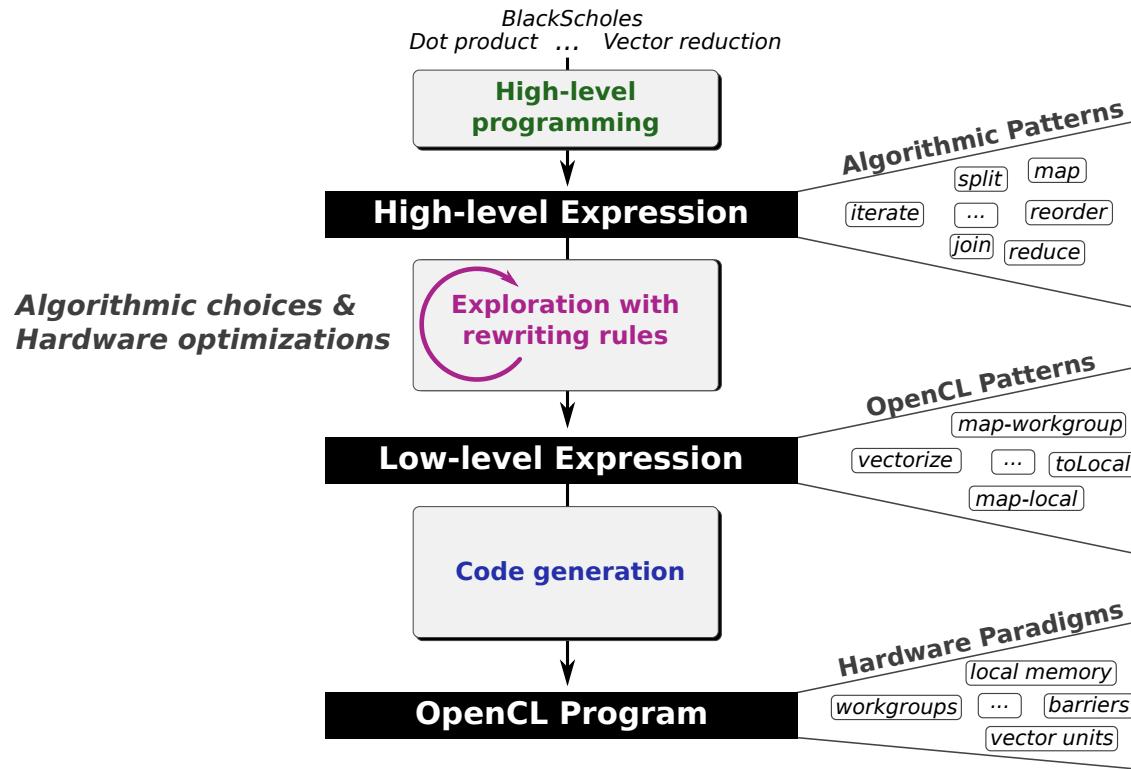
(b) AMD's HD 7970 GPU.



(c) Intel's E5530 dual-socket CPU.

- ... unfortunately, optimisations in OpenCL are not portable!
- **Challenge:** how to achieving portable performance?

Generating Performance Portable Code using Rewrite Rules



- **Ambition:** automatic generation of *Performance Portable* code

Walkthrough

① $\text{sum}(\text{vec}) = \text{reduce}(+, 0, \text{vec})$

rewrite rules code generation

②

```
vecSum = reduce ∘ join ∘ map-workgroup (
    join ∘ toGlobal (map-local (map-seq id)) ∘ split 1 ∘
    join ∘ map-warp (
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 1 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 2 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 4 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 8 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 16 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 32 ∘
    ) ∘ split 64 ∘
    join ∘ map-local (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 64 ∘
    join ∘ toLocal (map-local (reduce-seq (+) 0)) ∘
    split (blockSize/128) ∘ reorder-stride 128
) ∘ split blockSize
```

③

```
kernel
void reduce6(global float* g_idata,
             global float* g_odata,
             unsigned int n,
             local volatile float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i =
        get_group_id(0) * (get_local_size(0)*2)
        + get_local_id(0);
    unsigned int gridSize =
        WG_SIZE * get_num_groups(0);
    l_data[tid] = 0;
    while (i < n) {
        l_data[tid] += g_idata[i];
        if (i + WG_SIZE < n)
            l_data[tid] += g_idata[i+WG_SIZE];
        i += gridSize;
    }
    barrier(CLK_LOCAL_MEM_FENCE);

    if (WG_SIZE >= 256) {
        if (tid < 128) {
            l_data[tid] += l_data[tid+128];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (WG_SIZE >= 128) {
        if (tid < 64) {
            l_data[tid] += l_data[tid+ 64];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (tid < 32) {
        if (WG_SIZE >= 64) {
            l_data[tid] += l_data[tid+32];
        }
        if (WG_SIZE >= 32) {
            l_data[tid] += l_data[tid+16];
        }
        if (WG_SIZE >= 16) {
            l_data[tid] += l_data[tid+ 8];
        }
        if (WG_SIZE >= 8) {
            l_data[tid] += l_data[tid+ 4];
        }
        if (WG_SIZE >= 4) {
            l_data[tid] += l_data[tid+ 2];
        }
        if (WG_SIZE >= 2) {
            l_data[tid] += l_data[tid+ 1];
        }
    }
    if (tid == 0)
        g_odata[get_group_id(0)] = l_data[0];
}
```

Walkthrough

① $\text{sum}(\text{vec}) = \text{reduce}(+, 0, \text{vec})$

rewrite rules code generation

②

```
vecSum = reduce ∘ join ∘ map-workgroup (
    join ∘ toGlobal (map-local (map-seq id)) ∘ split 1 ∘
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        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 1 ∘
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        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 4 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 8 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 16 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 32 ∘
    ) ∘ split 64 ∘
    join ∘ map-local (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 64 ∘
    join ∘ toLocal (map-local (reduce-seq (+) 0)) ∘
    split (blockSize/128) ∘ reorder-stride 128 ∘
) ∘ split blockSize
```

③

```
kernel
void reduce6(global float* g_idata,
             global float* g_odata,
             unsigned int n,
             local volatile float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i =
        get_group_id(0) * (get_local_size(0)*2)
        + get_local_id(0);
    unsigned int gridSize =
        WG_SIZE * get_num_groups(0);
    l_data[tid] = 0;
    while (i < n) {
        l_data[tid] += g_idata[i];
        if (i + WG_SIZE < n)
            l_data[tid] += g_idata[i+WG_SIZE];
        i += gridSize;
    }
    barrier(CLK_LOCAL_MEM_FENCE);

    if (WG_SIZE >= 256) {
        if (tid < 128) {
            l_data[tid] += l_data[tid+128];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (WG_SIZE >= 128) {
        if (tid < 64) {
            l_data[tid] += l_data[tid+ 64];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (tid < 32) {
        if (WG_SIZE >= 64) {
            l_data[tid] += l_data[tid+32];
        }
        if (WG_SIZE >= 32) {
            l_data[tid] += l_data[tid+16];
        }
        if (WG_SIZE >= 16) {
            l_data[tid] += l_data[tid+ 8];
        }
        if (WG_SIZE >= 8) {
            l_data[tid] += l_data[tid+ 4];
        }
        if (WG_SIZE >= 4) {
            l_data[tid] += l_data[tid+ 2];
        }
        if (WG_SIZE >= 2) {
            l_data[tid] += l_data[tid+ 1];
        }
    }
    if (tid == 0)
        g_odata[get_group_id(0)] = l_data[0];
}
```

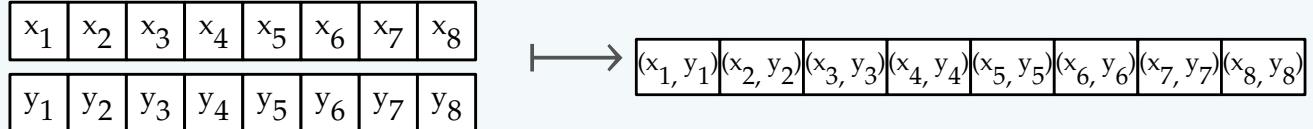


① Algorithmic Primitives (a.k.a. algorithmic skeletons)

$\text{map}(f, x)$:



$\text{zip}(x, y)$:



$\text{reduce}(+, 0, x)$:



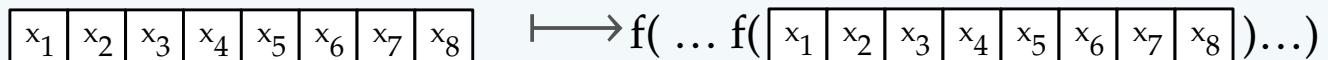
$\text{split}(n, x)$:



$\text{join}(x)$:



$\text{iterate}(f, n, x)$:



$\text{reorder}(\sigma, x)$:



① High-Level Programs

```
scal(a, vec) = map(λ x ↦ x*a, vec)
```

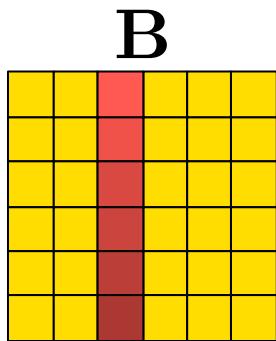
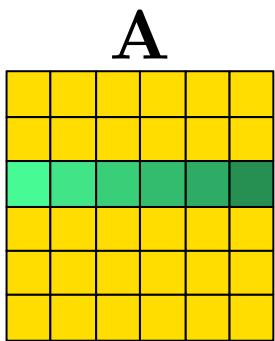
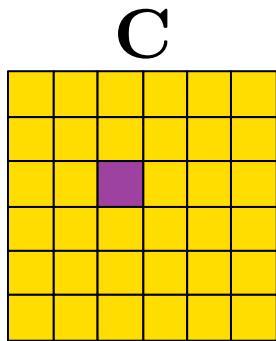
```
asum(vec) = reduce(+, 0, map(abs, vec))
```

```
dotProduct(x, y) = reduce(+, 0, map(*, zip(x, y)))
```

```
gemv(mat, x, y, α, β) =  
  map(+, zip(  
    map(λ row ↦ scal(α, dotProduct(row, x)), mat),  
    scal(β, y) ) )
```



① High-Level Programs



$A \times B =$

```
map(λ rowA ↦  
    map(λ colB ↦  
        dotProduct(rowA, colB)  
        , transpose(B))  
    , A)
```

Walkthrough

① $\text{sum}(\text{vec}) = \text{reduce}(+, 0, \text{vec})$

I
rewrite rules code generation

②

```
vecSum = reduce ∘ join ∘ map-workgroup (
    join ∘ toGlobal (map-local (map-seq id)) ∘ split 1 ∘
    join ∘ map-warp (
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 1 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 2 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 4 ∘
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        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 32 ∘
    ) ∘ split 64 ∘
    join ∘ map-local (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 64 ∘
    join ∘ toLocal (map-local (reduce-seq (+) 0)) ∘
    split (blockSize/128) ∘ reorder-stride 128
) ∘ split blockSize
```

③

```
kernel
void reduce6(global float* g_idata,
             global float* g_odata,
             unsigned int n,
             local volatile float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i =
        get_group_id(0) * (get_local_size(0)*2)
        + get_local_id(0);
    unsigned int gridSize =
        WG_SIZE * get_num_groups(0);
    l_data[tid] = 0;
    while (i < n) {
        l_data[tid] += g_idata[i];
        if (i + WG_SIZE < n)
            l_data[tid] += g_idata[i+WG_SIZE];
        i += gridSize;
    }
    barrier(CLK_LOCAL_MEM_FENCE);

    if (WG_SIZE >= 256) {
        if (tid < 128) {
            l_data[tid] += l_data[tid+128];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (WG_SIZE >= 128) {
        if (tid < 64) {
            l_data[tid] += l_data[tid+ 64];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
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    if (tid < 32) {
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        }
        if (WG_SIZE >= 32) {
            l_data[tid] += l_data[tid+16];
        }
        if (WG_SIZE >= 16) {
            l_data[tid] += l_data[tid+ 8];
        }
        if (WG_SIZE >= 8) {
            l_data[tid] += l_data[tid+ 4];
        }
        if (WG_SIZE >= 4) {
            l_data[tid] += l_data[tid+ 2];
        }
        if (WG_SIZE >= 2) {
            l_data[tid] += l_data[tid+ 1];
        }
    }
    if (tid == 0)
        g_odata[get_group_id(0)] = l_data[0];
}
```



Walkthrough

① $\text{sum}(\text{vec}) = \text{reduce}(+, 0, \text{vec})$

I
rewrite rules

code generation

②

```
vecSum = reduce ∘ join ∘ map-workgroup (
    join ∘ toGlobal (map-local (map-seq id)) ∘ split 1 ∘
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        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 1 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 2 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 4 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 8 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 16 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 32 ∘
    ) ∘ split 64 ∘
    join ∘ map-local (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 64 ∘
    join ∘ toLocal (map-local (reduce-seq (+) 0)) ∘
    split (blockSize/128) ∘ reorder-stride 128
) ∘ split blockSize
```

③

```
kernel
void reduce6(global float* g_idata,
             global float* g_odata,
             unsigned int n,
             local volatile float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i =
        get_group_id(0) * (get_local_size(0)*2)
        + get_local_id(0);
    unsigned int gridSize =
        WG_SIZE * get_num_groups(0);
    l_data[tid] = 0;
    while (i < n) {
        l_data[tid] += g_idata[i];
        if (i + WG_SIZE < n)
            l_data[tid] += g_idata[i+WG_SIZE];
        i += gridSize;
    }
    barrier(CLK_LOCAL_MEM_FENCE);

    if (WG_SIZE >= 256) {
        if (tid < 128) {
            l_data[tid] += l_data[tid+128];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (WG_SIZE >= 128) {
        if (tid < 64) {
            l_data[tid] += l_data[tid+ 64];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (tid < 32) {
        if (WG_SIZE >= 64) {
            l_data[tid] += l_data[tid+32];
        }
        if (WG_SIZE >= 32) {
            l_data[tid] += l_data[tid+16];
        }
        if (WG_SIZE >= 16) {
            l_data[tid] += l_data[tid+ 8];
        }
        if (WG_SIZE >= 8) {
            l_data[tid] += l_data[tid+ 4];
        }
        if (WG_SIZE >= 4) {
            l_data[tid] += l_data[tid+ 2];
        }
        if (WG_SIZE >= 2) {
            l_data[tid] += l_data[tid+ 1];
        }
    }
    if (tid == 0)
        g_odata[get_group_id(0)] = l_data[0];
}
```



② Algorithmic Rewrite Rules

- **Provably correct** rewrite rules
- Express algorithmic implementation choices

Split-join rule:

$$\text{map } f \rightarrow \text{join} \circ \text{map} (\text{map } f) \circ \text{split } n$$

Map fusion rule:

$$\text{map } f \circ \text{map } g \rightarrow \text{map} (f \circ g)$$

Reduce rules:

$$\text{reduce } f z \rightarrow \text{reduce } f z \circ \text{reducePart } f z$$

$$\text{reducePart } f z \rightarrow \text{reducePart } f z \circ \text{reorder}$$

$$\text{reducePart } f z \rightarrow \text{join} \circ \text{map} (\text{reducePart } f z) \circ \text{split } n$$

$$\text{reducePart } f z \rightarrow \text{iterate } n (\text{reducePart } f z)$$

② OpenCL Primitives

Primitive

mapGlobal

mapWorkgroup

mapLocal

mapSeq

reduceSeq

toLocal , *toGlobal*

mapVec,
splitVec, *joinVec*

OpenCL concept

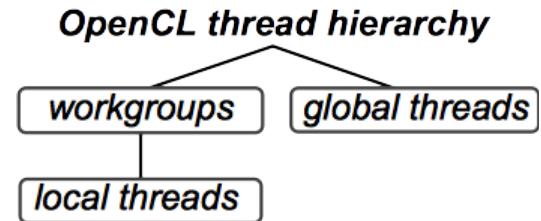
Work-items

Work-groups

Sequential implementations

Memory areas

Vectorisation



② OpenCL Rewrite Rules

- Express low-level implementation and optimisation choices

Map rules:

$$\text{map } f \rightarrow \text{mapWorkgroup } f \mid \text{mapLocal } f \mid \text{mapGlobal } f \mid \text{mapSeq } f$$

Local/ global memory rules:

$$\text{mapLocal } f \rightarrow \text{toLocal} (\text{mapLocal } f) \qquad \qquad \text{mapLocal } f \rightarrow \text{toGlobal} (\text{mapLocal } f)$$

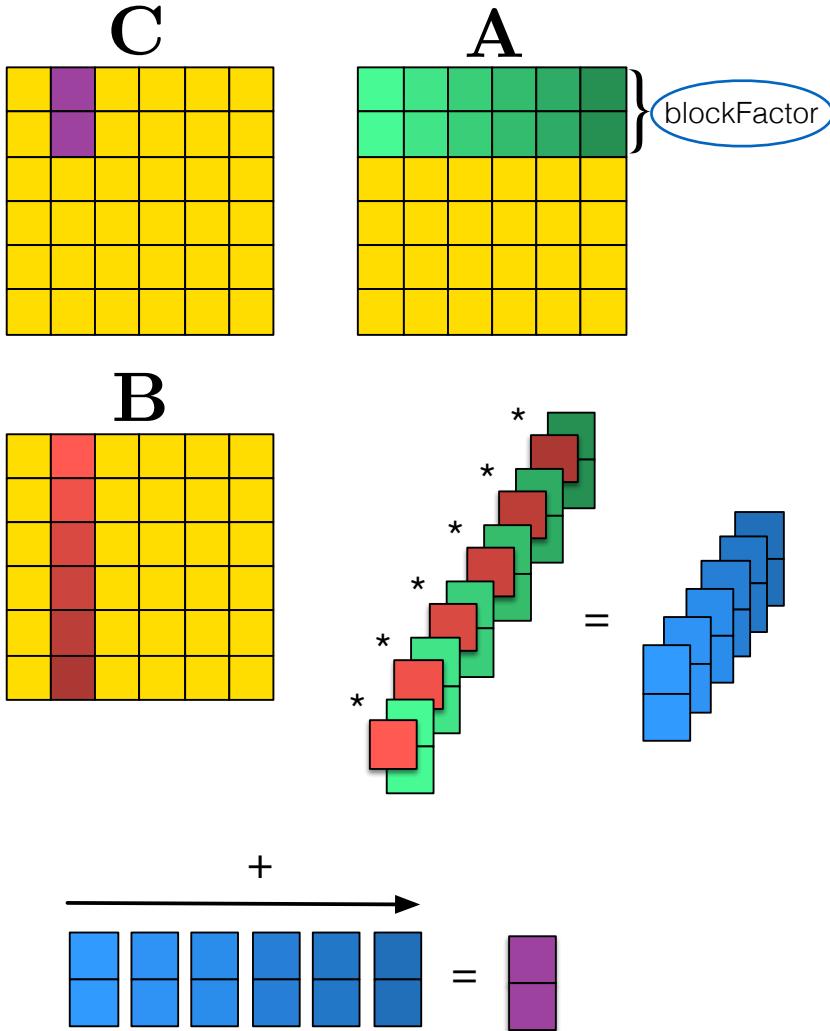
Vectorisation rule:

$$\text{map } f \rightarrow \text{joinVec} \circ \text{map} (\text{mapVec } f) \circ \text{splitVec } n$$

Fusion rule:

$$\text{reduceSeq } f \ z \circ \text{mapSeq } g \rightarrow \text{reduceSeq} (\lambda (acc, x). \ f (acc, g \ x)) \ z$$

② Optimisation Example: Register Blocking



```

1  kernel void KERNEL(
2      const global float* restrict A,
3      const global float* restrict B,
4      global float* C, int K, int M, int N)
5  {
6      float acc[blockFactor];
7
8      for (int glb_id_1 = get_global_id(1);
9          glb_id_1 < M / blockFactor;
10         glb_id_1 += get_global_size(1)) {
11          for (int glb_id_0 = get_global_id(0); glb_id_0 < N;
12              glb_id_0 += get_global_size(0)) {
13
14              for (int i = 0; i < K; i += 1)
15                  float temp = B[i * N + glb_id_0];
16                  for (int j = 0; j < blockFactor; j += 1)
17                      acc[j] +=
18                          A[blockFactor * glb_id_1 * K + j * K + i]
19                          * temp;
20
21              for (int j = 0; j < blockFactor; j += 1)
22                  C[blockFactor * glb_id_1 * N + j * N + glb_id_0]
23                  = acc[j];
24          }
25      }
26  }

```

② Register Blocking as a Macro Rule

- Optimisations are expressed as *Macro Rules*:
 - Series of Rewrites applied to achieve an optimisation goal

registerBlocking =

$\text{Map}(f) \Rightarrow \text{Join}() \circ \text{Map}(\text{Map}(f)) \circ \text{Split}(k)$

$\text{Map}(a \mapsto \text{Map}(b \mapsto f(a, b))) \Rightarrow \text{Transpose}() \circ \text{Map}(b \mapsto \text{Map}(a \mapsto f(a, b)))$

$\text{Map}(f \circ g) \Rightarrow \text{Map}(f) \circ \text{Map}(g)$

$\text{Map}(\text{Reduce}(f)) \Rightarrow \text{Transpose}() \circ \text{Reduce}((\text{acc}, x) \mapsto \text{Map}(f) \circ \text{Zip}(\text{acc}, x))$

$\text{Map}(\text{Map}(f)) \Rightarrow \text{Transpose}() \circ \text{Map}(\text{Map}(f)) \circ \text{Transpose}()$

$\text{Transpose}() \circ \text{Transpose}() \Rightarrow id$

$\text{Reduce}(f) \circ \text{Map}(g) \Rightarrow \text{Reduce}((\text{acc}, x) \mapsto f(\text{acc}, g(x)))$

$\text{Map}(f) \circ \text{Map}(g) \Rightarrow \text{Map}(f \circ g)$

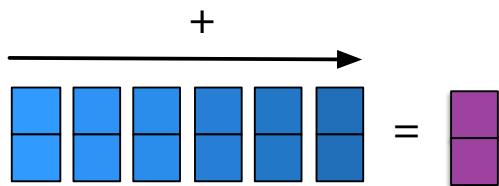
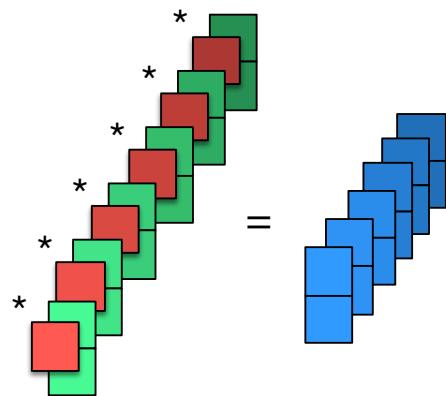
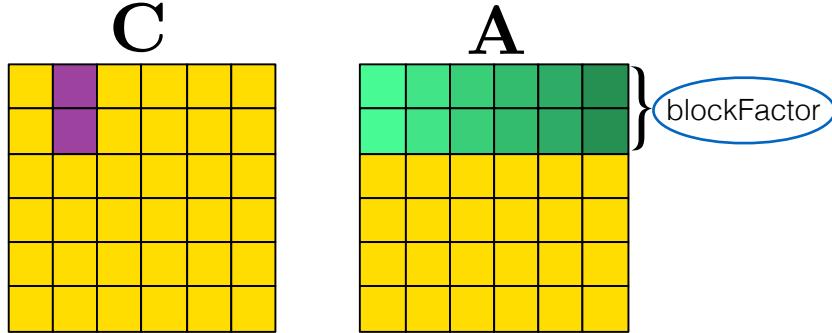
② Register Blocking as a Series of Rewrites

```
Map( $\overrightarrow{\text{rowA}}$   $\mapsto$   
Map( $\overrightarrow{\text{colB}}$   $\mapsto$   
Reduce(+)  $\circ$  Map(*)  
$ Zip( $\overrightarrow{\text{rowA}}, \overrightarrow{\text{colB}}$ )  
)  $\circ$  Transpose() $ B  
) $ A
```



```
Join()  $\circ$  Map(rowsA  $\mapsto$   
Transpose()  $\circ$  Map( $\overrightarrow{\text{colB}}$   $\mapsto$   
Transpose()  $\circ$  Reduce(( $\overrightarrow{\text{acc}}, \overrightarrow{\text{pair}}$ )  $\mapsto$   
Map( $x \mapsto x_{-0} + x_{-1} * \text{pair}_{-1}$ )  
$ Zip( $\overrightarrow{\text{acc}}, \text{pair}_{-0}$ )  
) $ Zip(Transpose() $ rowsA,  $\overrightarrow{\text{colB}}$ )  
)  $\circ$  Transpose() $ B  
)  $\circ$  Split(blockFactor) $ A
```

② Register Blocking Functionally Expressed



```


$$\begin{aligned}
& Join() \circ Map(rowsA \mapsto \\
& Transpose() \circ Map(\overrightarrow{colB} \mapsto \\
& Transpose() \circ Reduce((\overrightarrow{acc}, \overrightarrow{pair}) \mapsto \\
& \quad Map(x \mapsto x\_0 + x\_1 * pair\_1) \\
& \quad \$ Zip(\overrightarrow{acc}, pair\_0) \\
& ) \$ Zip(Transpose() \$ rowsA, \overrightarrow{colB}) \\
& ) \circ Transpose() \$ \mathbf{B} \\
& ) \circ Split(blockFactor) \$ \mathbf{A}
\end{aligned}$$


```

Walkthrough

① $\text{vecSum} = \text{reduce } (+) 0$

I
rewrite rules code generation

②

```
vecSum = reduce ∘ join ∘ map-workgroup (
    join ∘ toGlobal (map-local (map-seq id)) ∘ split 1 ∘
    join ∘ map-warp (
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 1 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 2 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 4 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 8 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 16 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 32 ∘
    ) ∘ split 64 ∘
    join ∘ map-local (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 64 ∘
    join ∘ toLocal (map-local (reduce-seq (+) 0)) ∘
    split (blockSize/128) ∘ reorder-stride 128
) ∘ split blockSize
```

③

```
kernel
void reduce6(global float* g_idata,
             global float* g_odata,
             unsigned int n,
             local volatile float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i =
        get_group_id(0) * (get_local_size(0)*2)
        + get_local_id(0);
    unsigned int gridSize =
        WG_SIZE * get_num_groups(0);
    l_data[tid] = 0;
    while (i < n) {
        l_data[tid] += g_idata[i];
        if (i + WG_SIZE < n)
            l_data[tid] += g_idata[i+WG_SIZE];
        i += gridSize;
    }
    barrier(CLK_LOCAL_MEM_FENCE);

    if (WG_SIZE >= 256) {
        if (tid < 128) {
            l_data[tid] += l_data[tid+128];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (WG_SIZE >= 128) {
        if (tid < 64) {
            l_data[tid] += l_data[tid+ 64];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (tid < 32) {
        if (WG_SIZE >= 64) {
            l_data[tid] += l_data[tid+32];
        }
        if (WG_SIZE >= 32) {
            l_data[tid] += l_data[tid+16];
        }
        if (WG_SIZE >= 16) {
            l_data[tid] += l_data[tid+ 8];
        }
        if (WG_SIZE >= 8) {
            l_data[tid] += l_data[tid+ 4];
        }
        if (WG_SIZE >= 4) {
            l_data[tid] += l_data[tid+ 2];
        }
        if (WG_SIZE >= 2) {
            l_data[tid] += l_data[tid+ 1];
        }
    }
    if (tid == 0)
        g_odata[get_group_id(0)] = l_data[0];
}
```

Walkthrough

①

$\text{vecSum} = \text{reduce } (+) \ 0$

rewrite rules

②

```
vecSum = reduce ∘ join ∘ map-workgroup (
    join ∘ toGlobal (map-local (map-seq id)) ∘ split 1 ∘
    join ∘ map-warp (
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 1 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 2 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 4 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 8 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 16 ∘
        join ∘ map-lane (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 32 ∘
    ) ∘ split 64 ∘
    join ∘ map-local (reduce-seq (+) 0) ∘ split 2 ∘ reorder-stride 64 ∘
    join ∘ toLocal (map-local (reduce-seq (+) 0)) ∘
    split (blockSize/128) ∘ reorder-stride 128 ∘
) ∘ split blockSize
```

③

code generation

```
kernel
void reduce6(global float* g_idata,
              global float* g_odata,
              unsigned int n,
              local volatile float* l_data) {
    unsigned int tid = get_local_id(0);
    unsigned int i =
        get_group_id(0) * (get_local_size(0)*2)
        + get_local_id(0);
    unsigned int gridSize =
        WG_SIZE * get_num_groups(0);
    l_data[tid] = 0;
    while (i < n) {
        l_data[tid] += g_idata[i];
        if (i + WG_SIZE < n)
            l_data[tid] += g_idata[i+WG_SIZE];
        i += gridSize;
    }
    barrier(CLK_LOCAL_MEM_FENCE);

    if (WG_SIZE >= 256) {
        if (tid < 128) {
            l_data[tid] += l_data[tid+128];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (WG_SIZE >= 128) {
        if (tid < 64) {
            l_data[tid] += l_data[tid+ 64];
        }
        barrier(CLK_LOCAL_MEM_FENCE);
    }
    if (tid < 32) {
        if (WG_SIZE >= 64) {
            l_data[tid] += l_data[tid+32];
        }
        if (WG_SIZE >= 32) {
            l_data[tid] += l_data[tid+16];
        }
        if (WG_SIZE >= 16) {
            l_data[tid] += l_data[tid+ 8];
        }
        if (WG_SIZE >= 8) {
            l_data[tid] += l_data[tid+ 4];
        }
        if (WG_SIZE >= 4) {
            l_data[tid] += l_data[tid+ 2];
        }
        if (WG_SIZE >= 2) {
            l_data[tid] += l_data[tid+ 1];
        }
    }
    if (tid == 0)
        g_odata[get_group_id(0)] = l_data[0];
}
```

③ Pattern based OpenCL Code Generation

- Generate OpenCL code for each OpenCL primitive

mapGlobal f xs →

```
for (int g_id = get_global_id(0); g_id < n;  
     g_id += get_global_size(0)) {  
    output[g_id] = f(xs[g_id]);  
}
```

reduceSeq f z xs →

```
T acc = z;  
for (int i = 0; i < n; ++i) {  
    acc = f(acc, xs[i]);  
}
```

⋮

⋮



Rewrite rules define a space of possible implementations

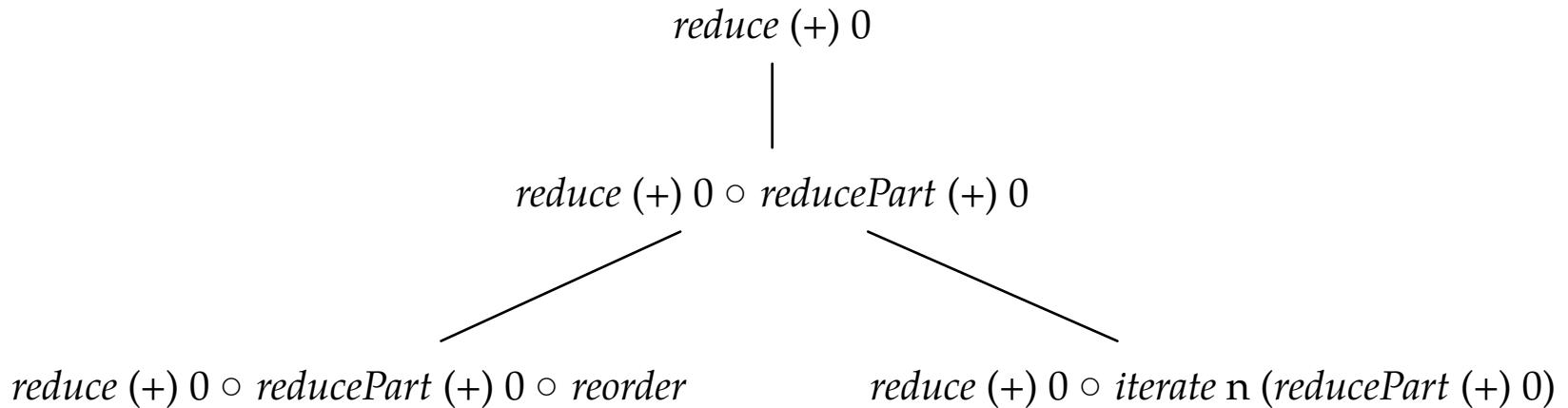
$$\begin{array}{c} \textit{reduce (+) 0} \\ | \\ \textit{reduce (+) 0} \circ \textit{reducePart (+) 0} \end{array}$$


Rewrite rules define a space of possible implementations

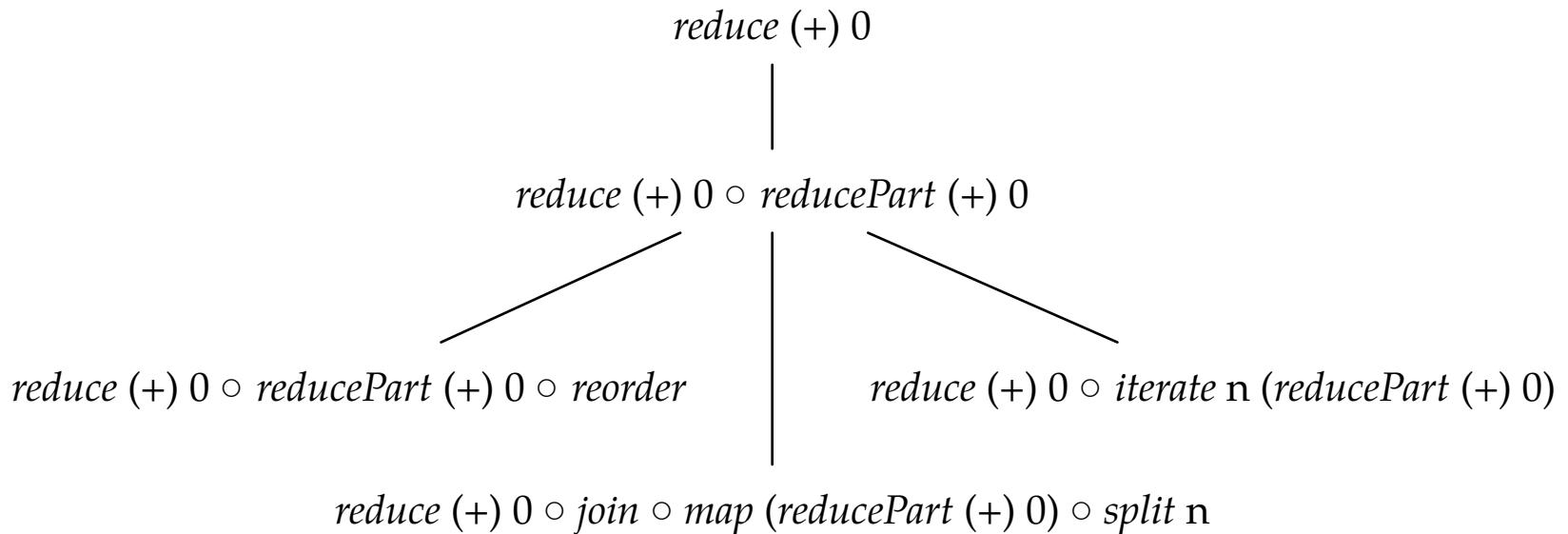
```
reduce (+) 0
|
reduce (+) 0 ○ reducePart (+) 0
/
reduce (+) 0 ○ reducePart (+) 0 ○ reorder
```



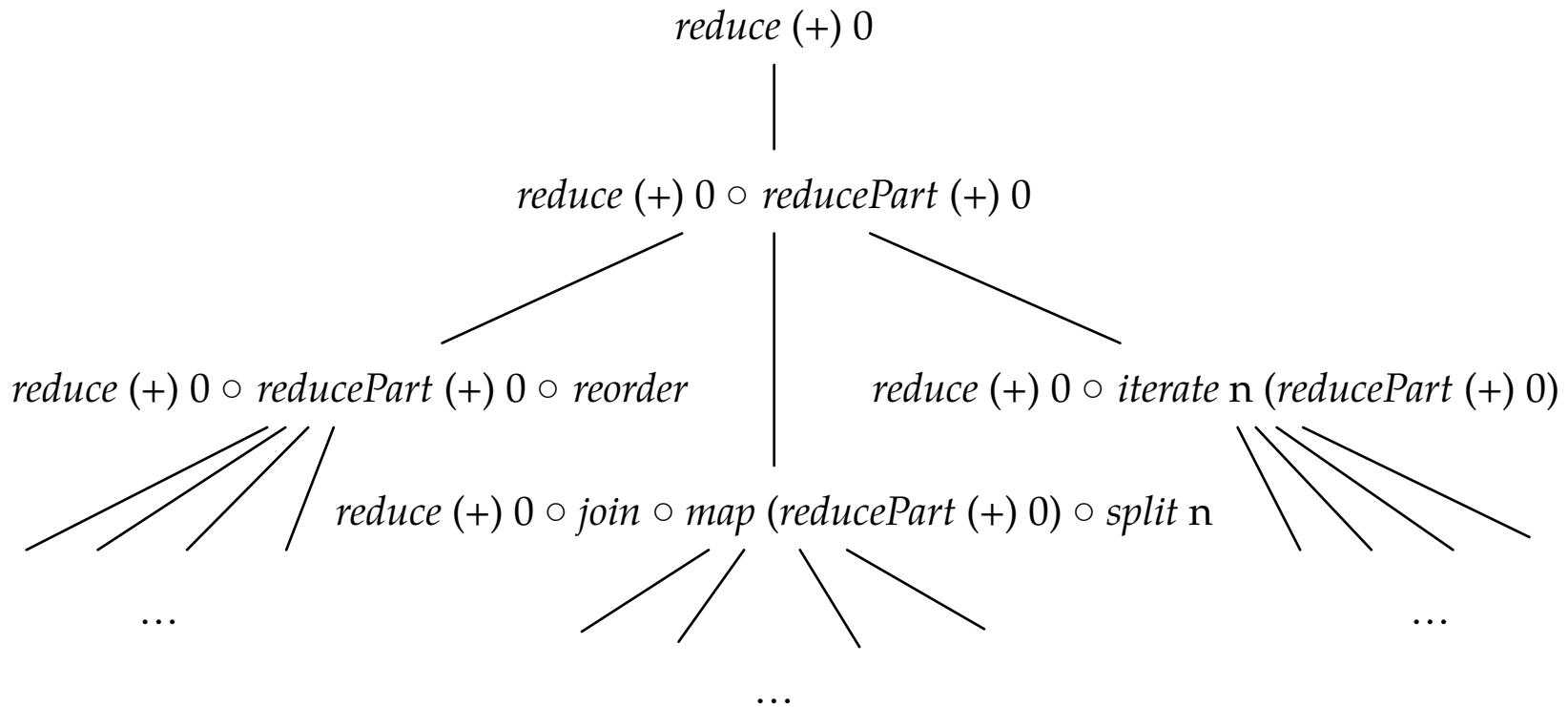
Rewrite rules define a space of possible implementations



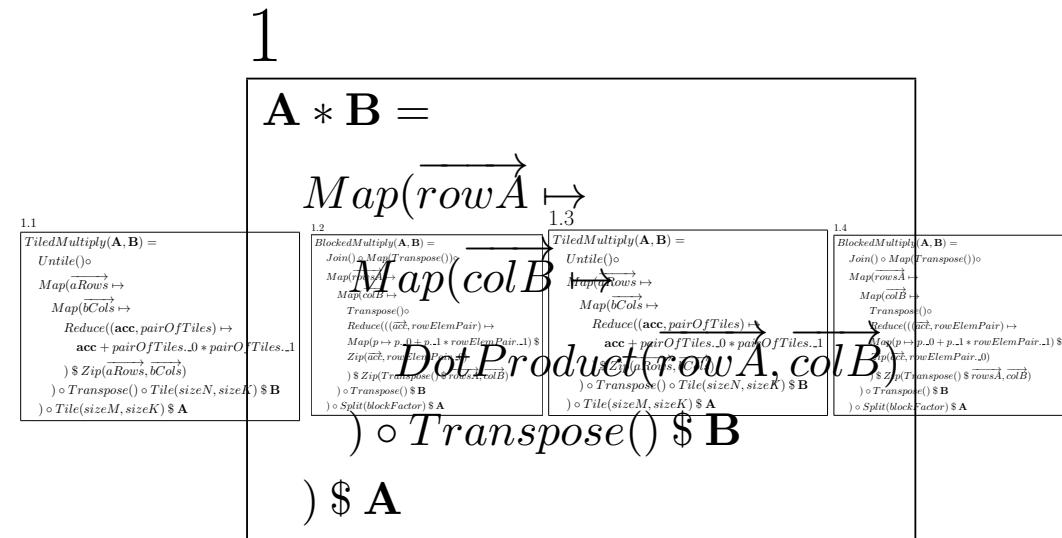
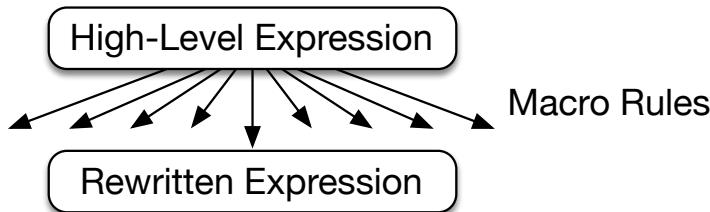
Rewrite rules define a space of possible implementations



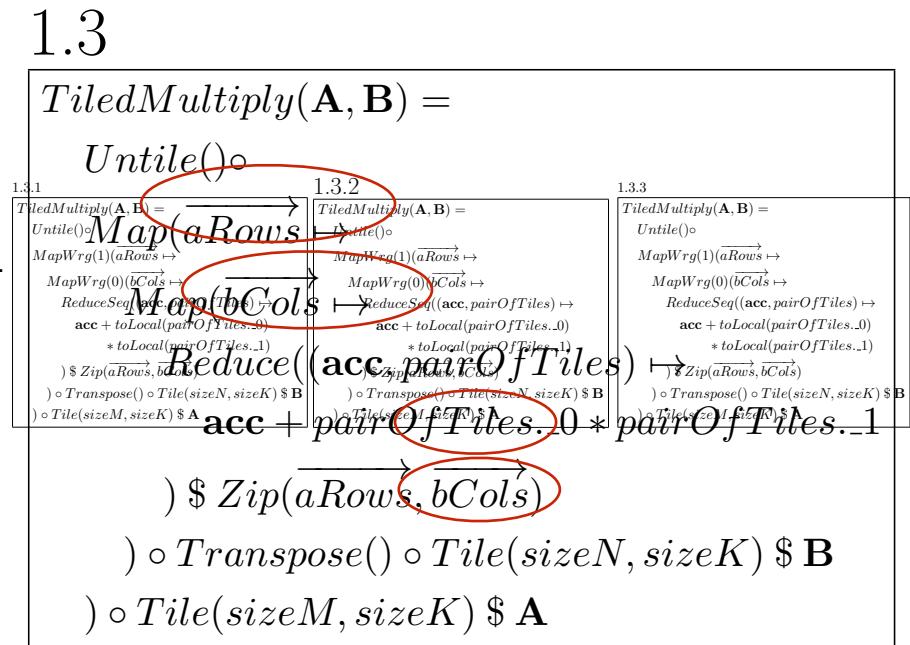
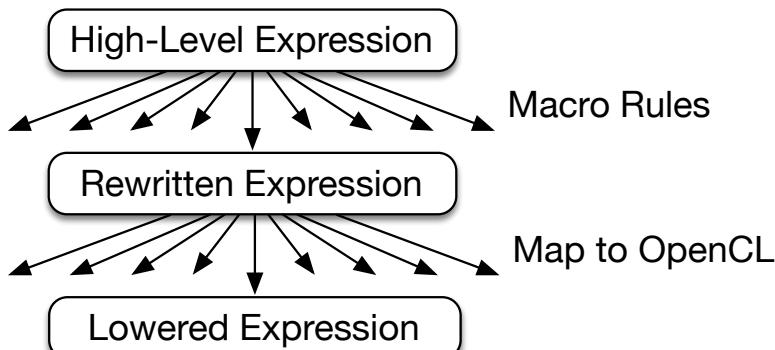
Rewrite rules define a space of possible implementations



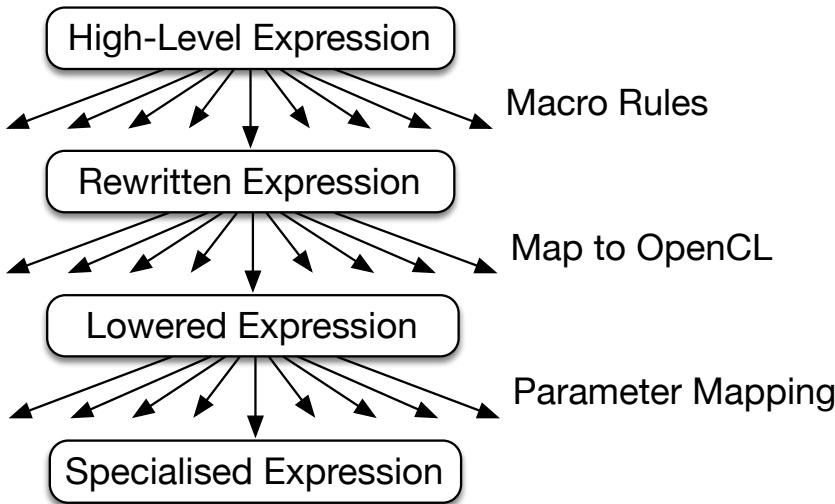
Exploration Strategy



Exploration Strategy



Exploration Strategy



1.3.2

$TiledMultiply(\mathbf{A}, \mathbf{B}) =$
 $Untile() \circ$

1.3.2.1 $MapWrg(1)(\overrightarrow{aRows} \mapsto$

$TiledMultiply(\mathbf{A}, \mathbf{B}) =$
 $Untile() \circ$

$MapWrg(1)(\overrightarrow{aRows} \mapsto$

$MapWrg(0)(\overrightarrow{bCols} \mapsto$

$ReduceSeq((acc, pairOfTiles_{..0}) \mapsto$
 $acc + toLocal(pairOfTiles_{..0})$
 $* toLocal(pairOfTiles_{..1})$
 $) \$ Zip(\overrightarrow{aRows}, \overrightarrow{bCols})$
 $) \circ Transpose() \circ Tile(128, 16) \$ \mathbf{B}$
 $) \circ Tile(128, 16) \$ \mathbf{A}$

1.3.2.4 $TiledMultiply(\mathbf{A}, \mathbf{B}) =$

$Untile() \circ$

$MapWrg(1)(\overrightarrow{aRows} \mapsto$

$\$ Zip(\overrightarrow{aRows}, \overrightarrow{bCols})$

$MapWrg(0)(\overrightarrow{bCols} \mapsto$

$ReduceSeq((acc, pairOfTiles_{..0}) \mapsto$
 $acc + toLocal(pairOfTiles_{..0})$
 $* toLocal(pairOfTiles_{..1})$
 $) \$ Zip(\overrightarrow{aRows}, \overrightarrow{bCols})$
 $) \circ Transpose() \circ Tile(128, 16) \$ \mathbf{B}$
 $) \circ Tile(128, 16) \$ \mathbf{A}$

1.3.2.5 $TiledMultiply(\mathbf{A}, \mathbf{B}) =$

$Untile() \circ$

$MapWrg(1)(\overrightarrow{aRows} \mapsto$

$\$ Zip(\overrightarrow{aRows}, \overrightarrow{bCols})$

$MapWrg(0)(\overrightarrow{bCols} \mapsto$

$ReduceSeq((acc, pairOfTiles_{..0}) \mapsto$
 $acc + toLocal(pairOfTiles_{..0})$
 $* toLocal(pairOfTiles_{..1})$
 $) \$ Zip(\overrightarrow{aRows}, \overrightarrow{bCols})$
 $) \circ Transpose() \circ Tile(128, 16) \$ \mathbf{B}$
 $) \circ Tile(128, 16) \$ \mathbf{A}$

1.3.2.6 $TiledMultiply(\mathbf{A}, \mathbf{B}) =$

$Untile() \circ$

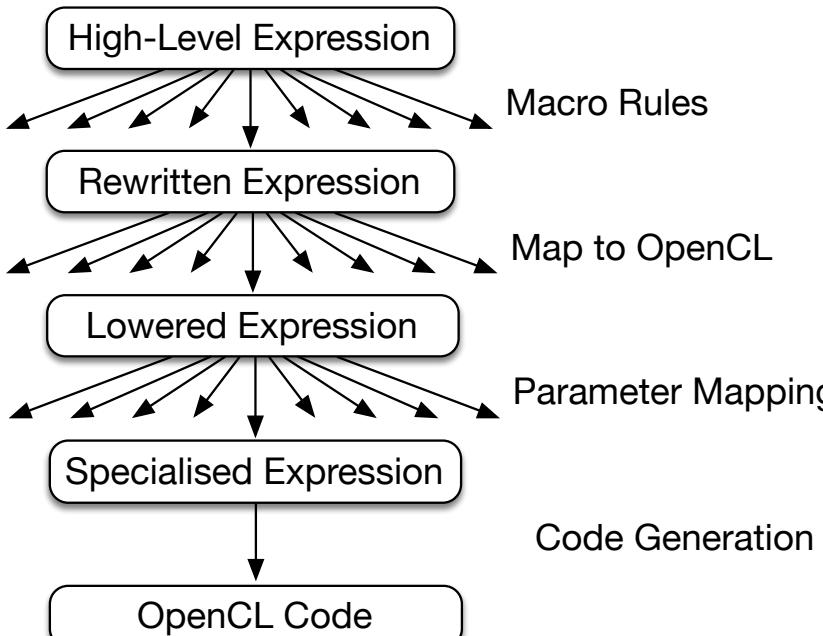
$MapWrg(1)(\overrightarrow{aRows} \mapsto$

$MapWrg(0)(\overrightarrow{bCols} \mapsto$

$ReduceSeq((acc, pairOfTiles_{..0}) \mapsto$
 $acc + toLocal(pairOfTiles_{..0})$
 $* toLocal(pairOfTiles_{..1})$
 $) \$ Zip(\overrightarrow{aRows}, \overrightarrow{bCols})$
 $) \circ Transpose() \circ Tile(128, 16) \$ \mathbf{B}$
 $) \circ Tile(128, 16) \$ \mathbf{A}$



Exploration Strategy



1.3.2.5

```

1 kernel mm_and_opt(global float *A, B,C,
2   int aRows, int aCols, int bRows, int bCols);
3   local float tileA[512]; tileB[512];
4
5 private float acc_0; ...; acc_31;
6 private float blockOfA_0; ...; blockOfA_7;
7 private float blockOfB_0; ...; blockOfB_7;
8
9 int lid0 = local_id(0); lid1 = local_id(1);
10 int wid0 = group_id(0)*wid + group_id(1);
11
12 for (int w1=wid1; w1<M/64; w1+=num_grps(1)) {
13   for (int w0=wid0; w0<N/64; w0+=num_grps(0)) {
14     acc_0 = 0.0f; acc_1 = 0.0f;
15     for (int i=0; i<K/8; i++) {
16       vstore4(vload4(lid1*M/4+2*i*M+16*w1+lid0,A), 16*lid1+lid0, tileA);
17       vstore4(vload4(lid1*M/4+2*i*M+N/16*w0+lid0,B), 16*lid1+lid0, tileB);
18       barrier (...);
19     }
20   }
21   barrier (...);
22   blockOfA_0 = tileA[0+lid1*N+lid0*8]; blockOfB_0 = tileB[0+lid1*N+lid0*8];
23   blockOfB_0 = tileB[0+64*j+lid0]; ...; blockOfB_3 = tileB[8+64*j+lid0];
24
25   acc_0 += blockOfA_0 * blockOfB_0; ...; acc_28 += blockOfA_7 * blockOfB_0;
26   acc_1 += blockOfA_0 * blockOfB_1; ...; acc_29 += blockOfA_7 * blockOfB_1;
27   acc_2 += blockOfA_0 * blockOfB_2; ...; acc_30 += blockOfA_7 * blockOfB_2;
28   acc_3 += blockOfA_0 * blockOfB_3; ...; acc_31 += blockOfA_7 * blockOfB_3;
29 }
30 barrier (...);
31 } $ Zip(aRows, bCols)
32
33 C[ 0+8*lid1*N+64*w0+64*w1*N+0*N+lid0]=acc_0; ...; C[ 0+8*lid1*N+64*w0+64*w1*N+7*N+lid0]=acc_28;
34 C[16+8*lid1*N+14*w0+64*w1*N+8*N+lid0]=acc_1; ...; C[16+8*lid1*N+14*w0+4*w1*N+7*N+lid0]=acc_29;
35 C[32+8*lid1*N+64*w0+64*w1*N+9*N+lid0]=acc_2; ...; C[32+8*lid1*N+64*w0+64*w1*N+7*N+lid0]=acc_30;
36 C[48+8*lid1*N+64*w0+64*w1*N+0*N+lid0]=acc_3; ...; C[48+8*lid1*N+64*w0+64*w1*N+7*N+lid0]=acc_31;
37 } } } ) $ Tile(128, 16) $ A
  
```

The code snippet shows the generated OpenCL kernel for tiled matrix multiplication. It defines a kernel *TiledMatrixMul(A, B)* that takes *aRows* and *bCols* as parameters. The kernel uses local memory tiles for *A* and *B* matrices. It iterates over *w1* and *w0* dimensions, applying macro rules to map the high-level expression to this specific implementation. The code includes parameter mapping for tile indices and accumulators, and concludes with a *Zip* operation followed by a *Tile* operation.



Heuristics for Matrix Multiplication

For Macro Rules:

- Nesting depth
- Distance of addition and multiplication
- Number of times rules are applied

For Map to OpenCL:

- Fixed parallelism mapping
- Limited choices for mapping to local and global memory
- Follows best practice

For Parameter Mapping:

- Amount of memory used
 - Global
 - Local
 - Registers
- Amount of parallelism
 - Work-items
 - Workgroup



Exploration in Numbers for Matrix Multiplication

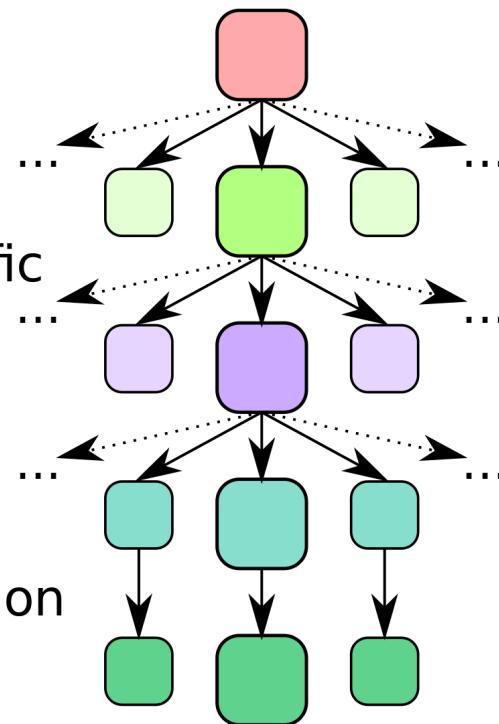
Phases:

Algorithmic Exploration

OpenCL specific Exploration

Parameter Exploration

Code Generation



Program Variants:

High-Level Program 1

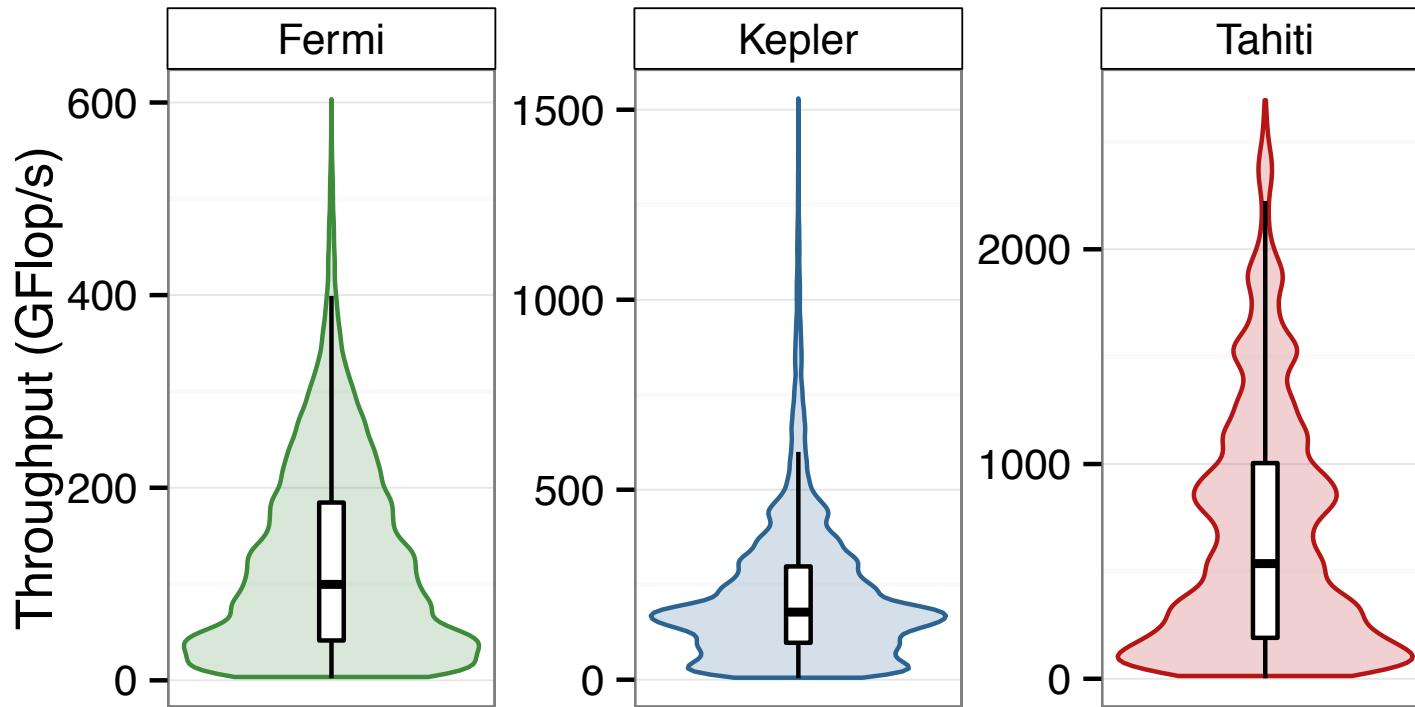
Algorithmic Rewritten Program 8

OpenCL Specific Program 760

Fully Specialized Program 46,000

OpenCL Code 46,000

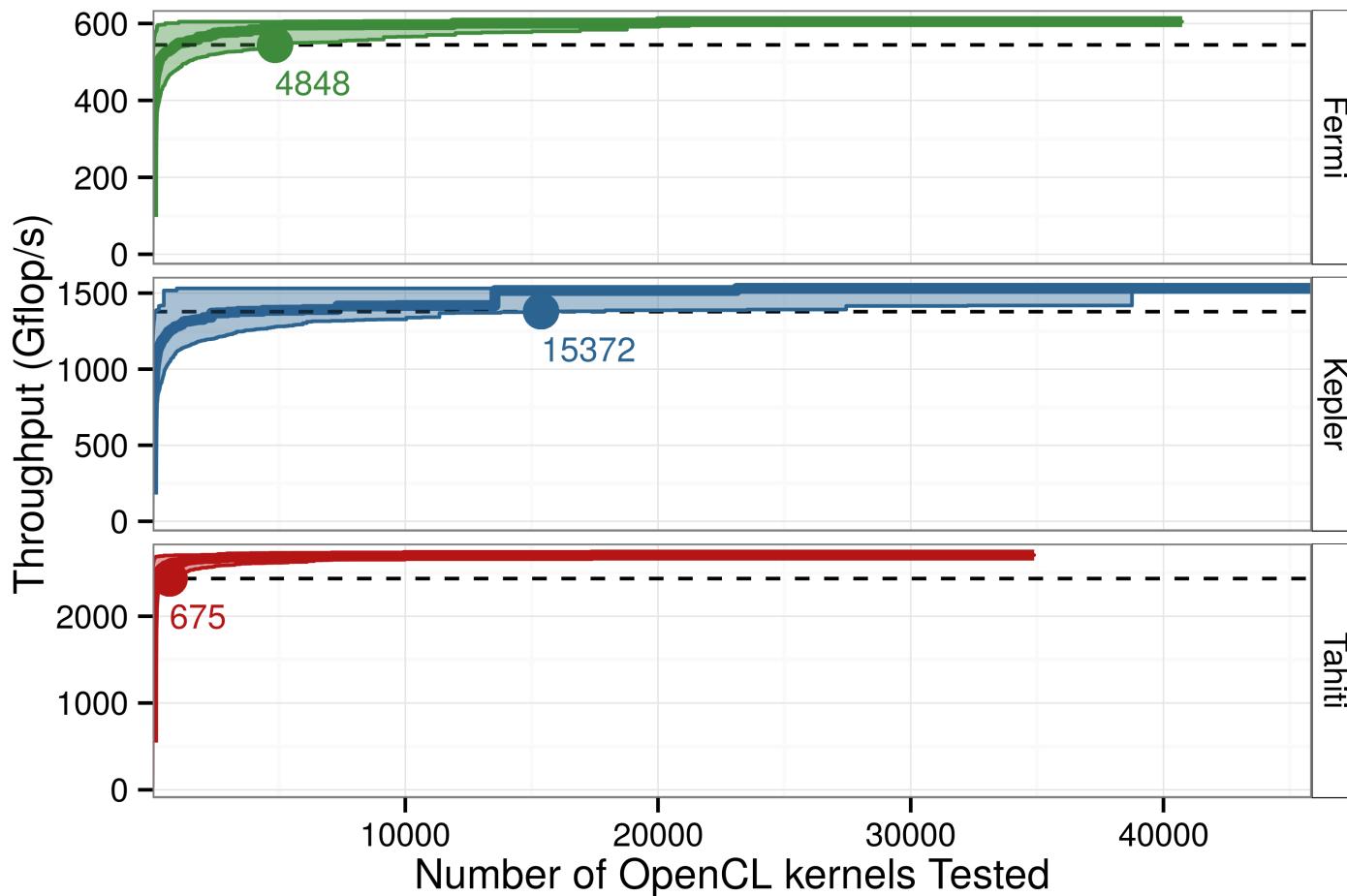
Exploration Space for Matrix Multiplication



Only few OpenCL kernel with very good performance

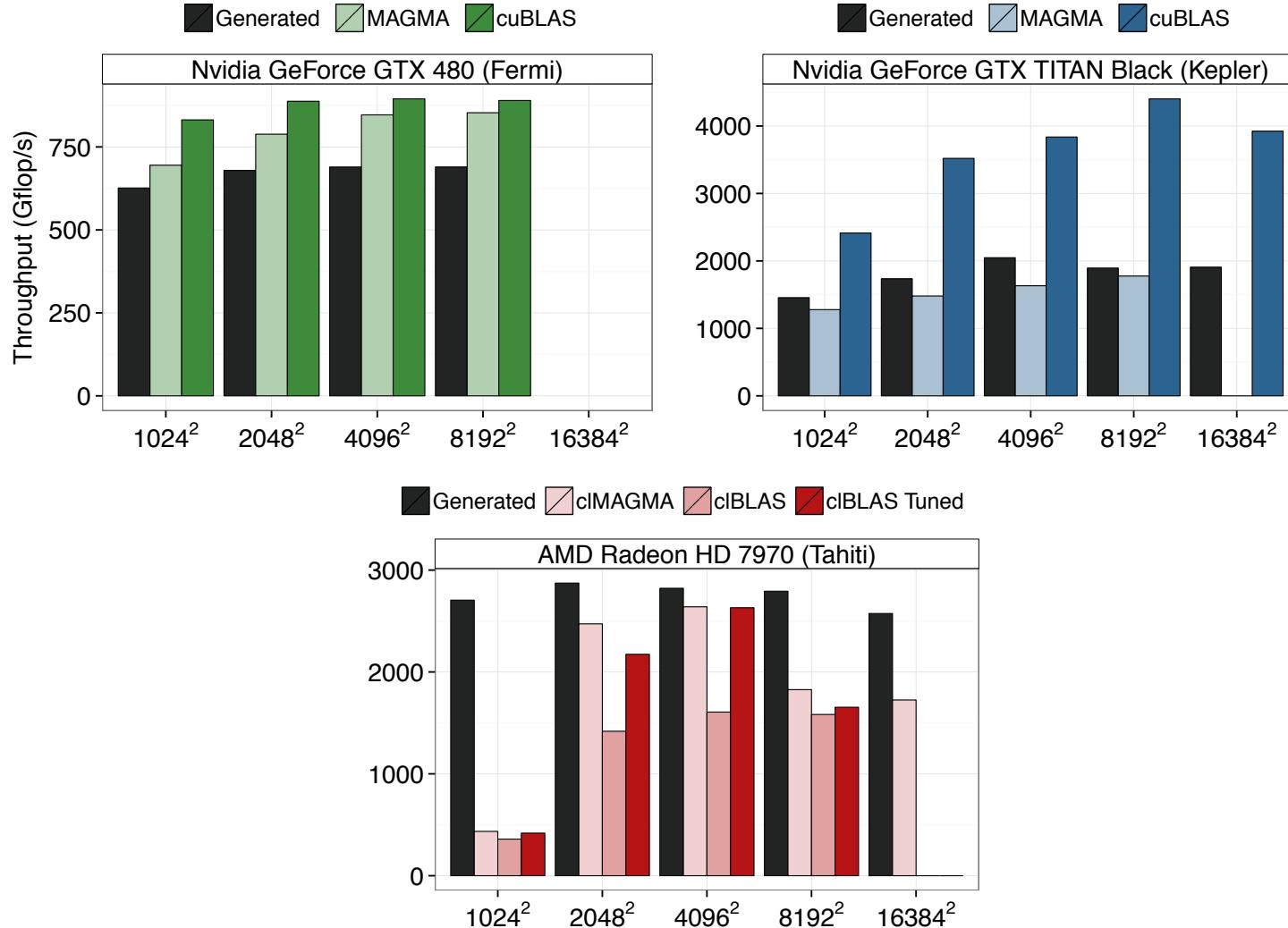


Performance Evolution for Randomised Search



Even with a simple random search strategy one can expect to find a good performing kernel quickly

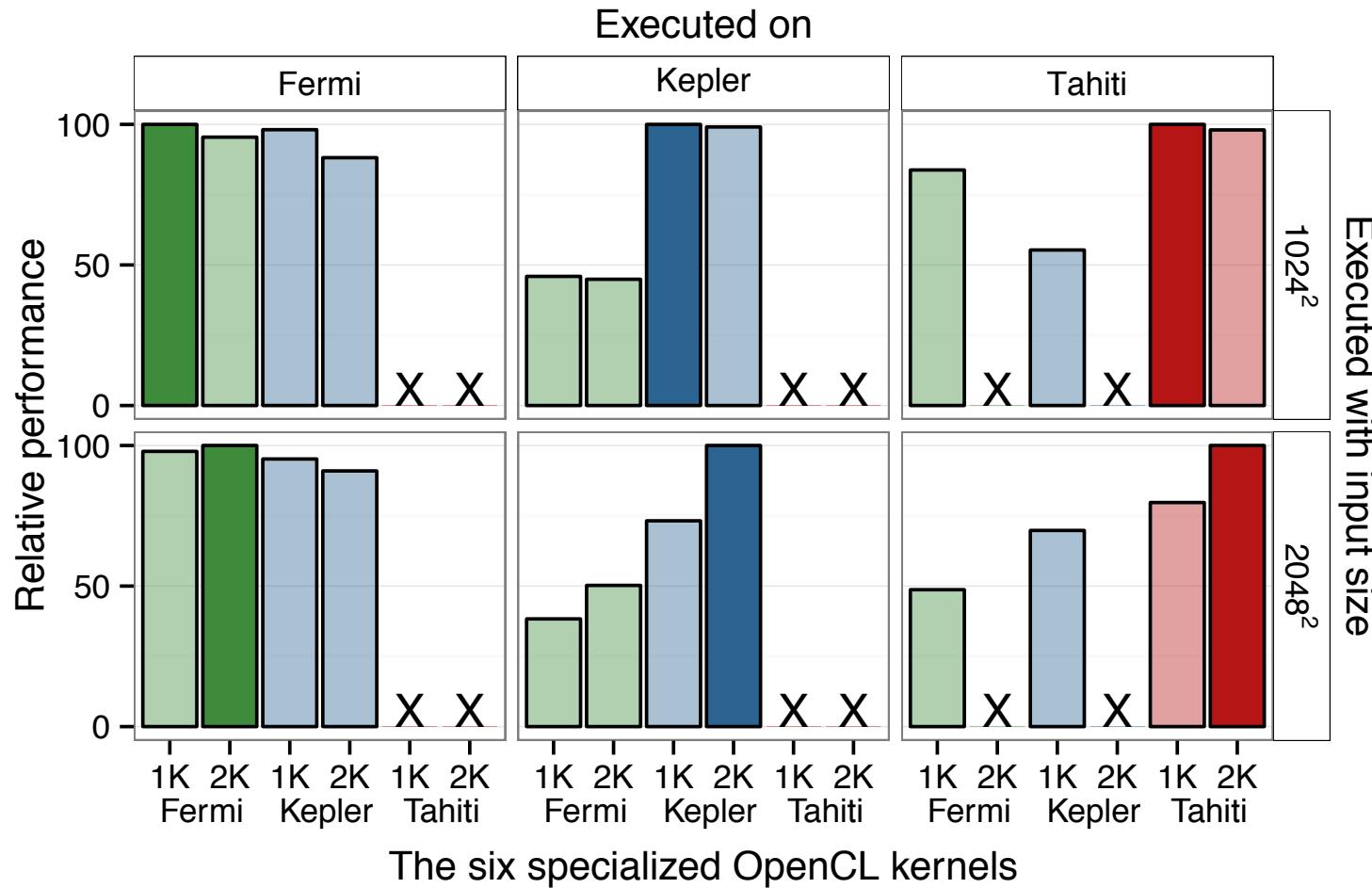
Performance Results Matrix Multiplication



Performance close or better than hand-tuned MAGMA library



Performance Portability Matrix Multiplication



Generated kernels are specialised for device and input size

Summary

- OpenCL code is *hard to write* and not *performance portable*
- Our approach uses
 - *portable* and functional **high-level primitives**,
 - **OpenCL-specific low-level primitives**, and
 - **rewrite-rules** to generate high *performance* code.
- Rewrite-rules define a space of possible implementations
- Performance on par with specialised, highly-tuned code



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Thibaut Lutz
Now with Nvidia



Toomas Remmelg
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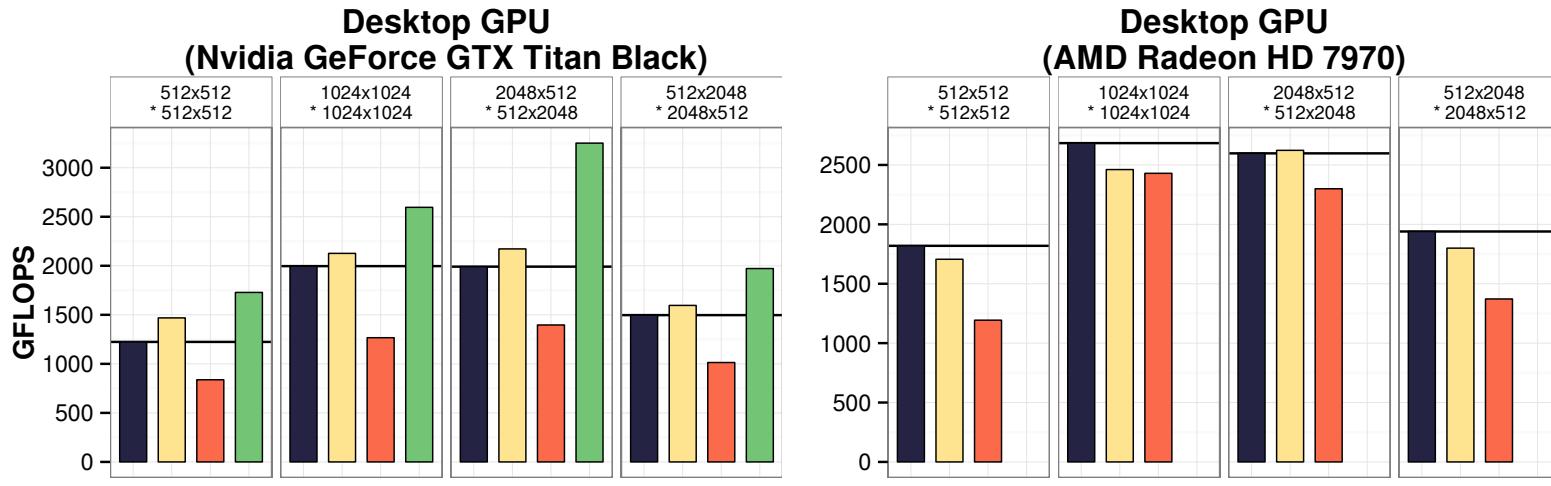
More details in the **ICFP 2015**, **GPGPU 2016**, **CASES 2016** papers available at:
<http://www.lift-project.org>

supported by:



Oracle Labs

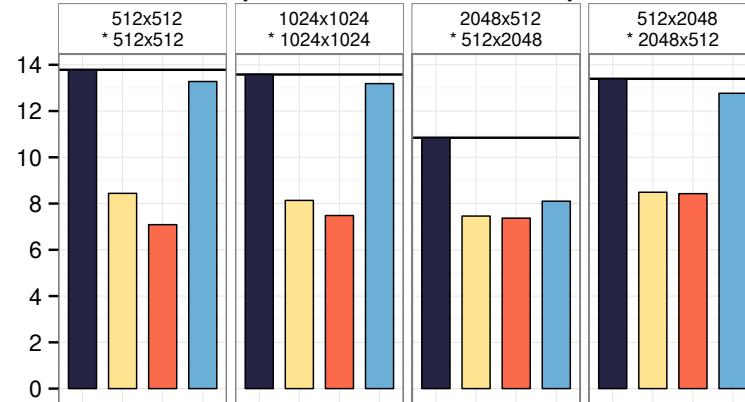
More Results Matrix Multiplication



Legend: Rewrite-based (dark blue), CLBlast + CLTune (yellow), cIBLAS (orange), cuBLAS (green)

Legend: Rewrite-based (dark blue), CLBlast + CLTune (yellow), cIBLAS (orange)

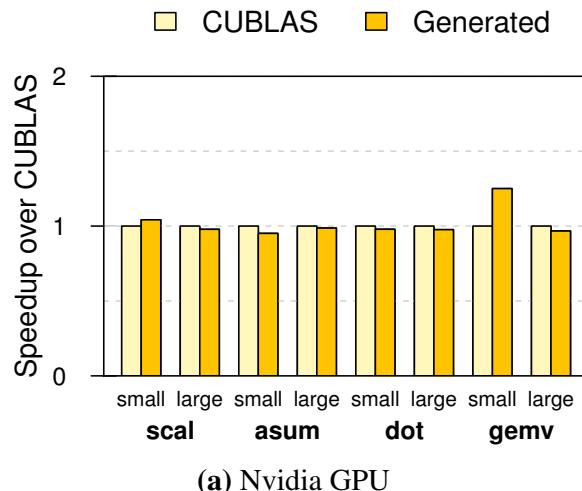
**Mobile GPU
(ARM Mali-T628 MP6)**



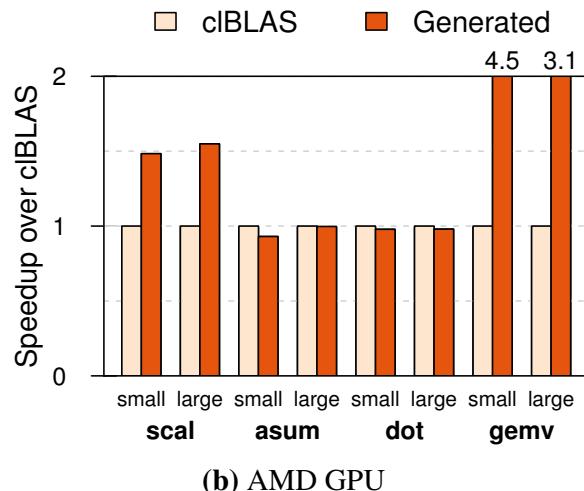
Legend: Rewrite-based (dark blue), CLBlast + CLTune (yellow), cIBLAS (orange), Hand optimized (light blue)



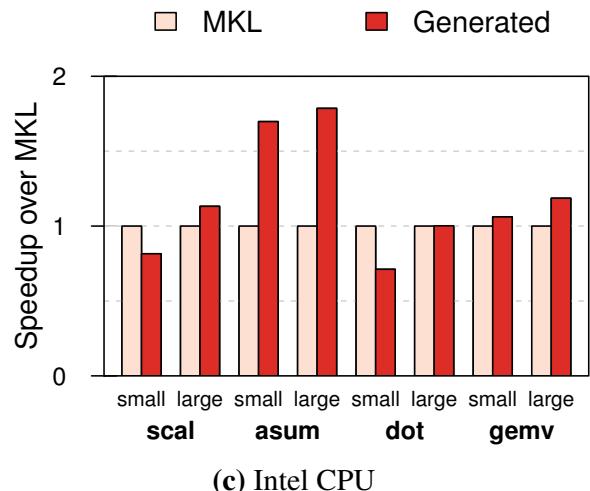
Performance Results more Benchmarks vs. Hardware-Specific Implementations



(a) Nvidia GPU



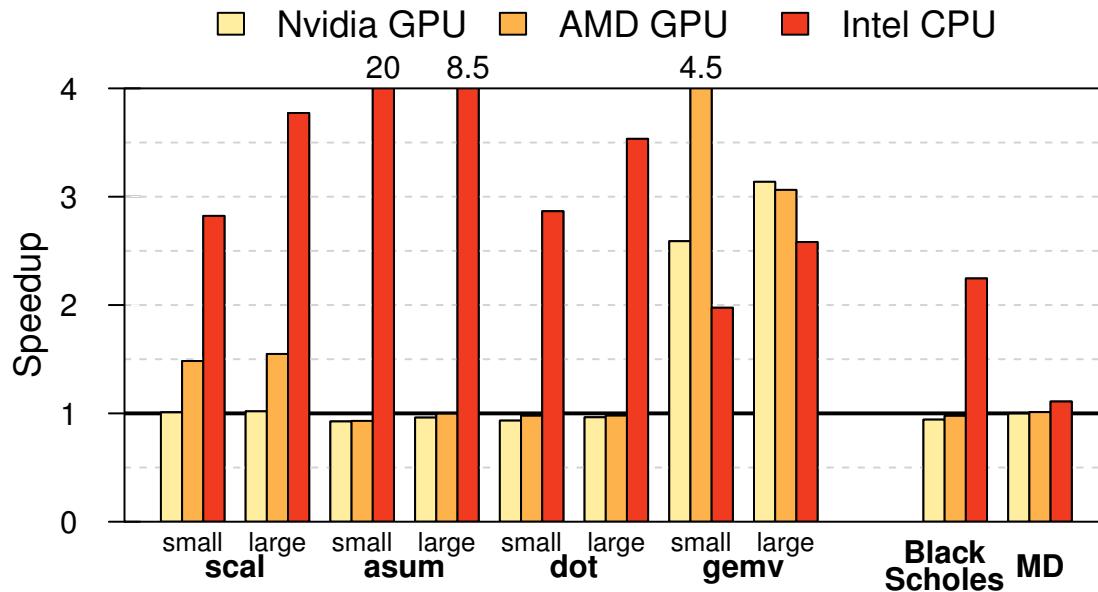
(b) AMD GPU



(c) Intel CPU

- Automatically generated code vs. expert written code
- Competitive performance vs. highly optimised implementations
- Up to **4.5x** speedup for *gemv* on AMD

Performance Results more Benchmarks vs. Portable Implementation



- Up to 20x speedup on fairly simple benchmarks vs. portable clBLAS implementation