

Rapid Customization of Image Processors Using Halide

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Motivation

- Exploit custom operations in computing platform at high abstraction level description
 - avoid error-prune low-level descriptions
 - maintain platform portability
- We demonstrate an dxperimental design flow
 - High-abstraction level descriptions: Halide
 - Processor customization: TCE toolset



Halide (MIT)

- Domain specific (image processing) functional parallel programming language developed at MIT
- Decoupling of algorithm and its schedule
 - The same algorithm easily optimized for different types of processors, only by modifying the schedule
- Algorithm part requires only little knowledge about parallel programming or parallel compute platforms



C/C++ vs. Halide

C++ function for "blur"

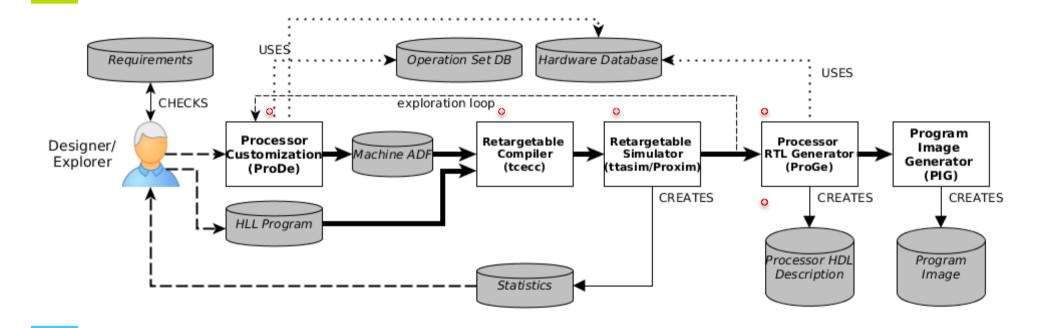
```
void box filter 3x3(const Image &in, Image &blury) {
   m128i one third = mm set1 epi16(21846);
  #pragma omp parallel for
  for (int yTile = 0; yTile < in.height(); yTile += 32) {</pre>
     m128i a, b, c, sum, avg;
     m128i blurx[(256/8)*(32+2)]; // allocate blurx array
   for (int xTile = 0; xTile < in.width; xTile += 256) {</pre>
       m128i *blurxPtr = blurx;
      for (int y = -1; y < 32+1; y++) {
        const uint16 t *inPtr = &(in[yTile+y][xTile]);
        for (int x = 0; x < 256; x += 8) {
          a = mm loadu si128(( mm128*)(inPtr-1));
          b = mm \log sil28((mm128*)(inPtr+1));
          c = mm \log sil28((mm128*)(inPtr));
          sum = mm add epi16( mm add epi16(a, b), c);
          avg = mm mulhi epi16(sum, one third);
          mm store sil28(blurxPtr++, avg);
          intPtr += 8;
     }}
      blurxPtr = blurx;
      for (int y = 0; y < 32; y++) {
         128i *outPtr = ( m128i *)(&(blury[yTile+y][xTile]));
        for (int x = 0; x < 256; x += 8) {
          a = mm load si128(blurxPtr+(2*256)/8);
          a = mm \log sil28(blurxPtr+256/8);
          a = mm load si128(blurxPtr++);
          sum = mm add epi16( mm add epi16(1, b), c);
          avg = mm mulhi epi16(sum, one third);
          mm store sil28(outPtr++, avg);
```

Halide description for "blur"

}}}}



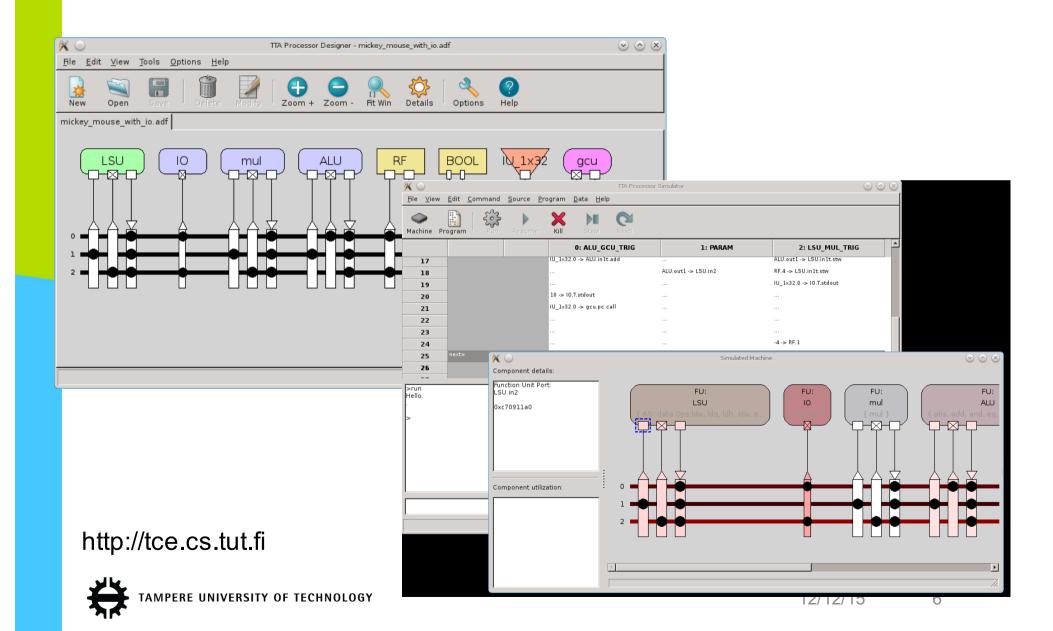
TTA-Based Co-Design Environment (TCE) (Tampere Univ. Tech.)



- Framework for customizing processor architectures and retargetable compiler
 - Exploits transport triggered architecture (TTA) template
 - Supports C/C++ and OpenCL
- A research platform for customized processors, compilation techniques etc.
- MIT-licensed, available at http://tce.cs.tut.fi



TCE Screenshots



Custom Operations

- Custom operation (special instruction) is an optimized atomic operation that usually wraps up the behavior of multiple basic operations.
- Simple custom operations automatically invoked by compiler (LLVM)
 - E.g., multiply-accumulate, auto-increment in addressing



Custom Operations

- More complex custom operations need to be created
 - Custom op created as a simulation model in TCE (for implementation the corresponding HDL has to be created)
 - TCE generates intrinsics for the custom op and corresponding function wrapper for Halide
 - User indicates the use of custom op with extern declaration

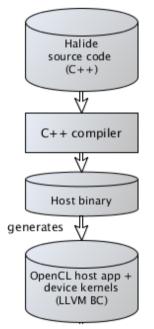


Halide-OpenCL Flow

Halide:

- Application compiled to executable binary for the target
- Application wraps the algorithm and the schedule into an OpenCL application presented in LLVM IR (OpenCL kernels included as global strings)
- Compiles LLVM IR for the target

(offline and online compilation possible)

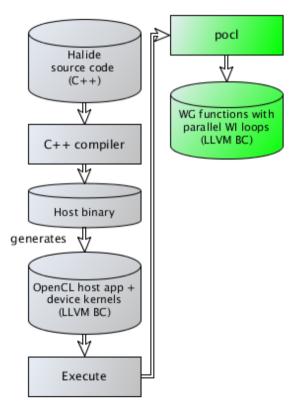




Halide-OpenCL-TCE Flow

OpenCL:

- OpenCL application uses pocl through the OpenCL API
 - pocl is our open source implementation of OpenCL standard: http://pocl.sourceforge.net/
- OpenCL takes care of platform portability, parallelism on higher level and heterogeneity.
- pocl compiles the plain text kernels to Workgroup functions
- pocl gathers execution times of the kernels for profiling purposes

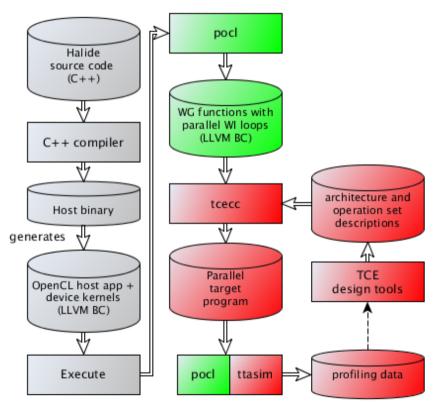




Halide-OpenCL-TCE Flow

TCE:

- Processor design tools
- Custom operation definitions
- Simulation models for custom ops
- Function wrapping for custom ops
- Retargetable compiler
 - adapts to changes in the processor architecture on the fly
- Profiling tools for evaluating performance





Experiment Case 1: Blur

- Blurs image by calculating weighted average over adjacent pixels
- Starting point
 - minimalistic scalar integer TTA processor
- Accelerated with a custom op
 - weighted average



Case 1: Blur

```
// Take a color 8-bit input
Image<uint8 t> input = load<uint8 t>("rgb.png");
// Upgrade it to 16-bit, so we can do math without it overflowing.
Func input 16("input 16");
input 16(x, y, c) = cast<uint16 t>(input(x, y, c));
// Blur it horizontally:
Func blur x("blur x");
blur_x(x, y, c) = (input_16(x-1, y, c)
+ 2*input_16(x, y, c) Potential custom operation
("*2" and "/4" are only rewiring
                      + input_16(x+1, y, c))/4; inside custom op)
// Blur it vertically:
Func blur y("blur y");
blur_y(x, y, c) = (blur_x(x, y-1, c)
+ 2*blur_x(x, y, c)
+ blur_x(x, y+1, c))/4;
                                                    Potential custom operation
```

Weighted average custom op

```
HalideExtern 3 (uint8 t, tce wavg3, uint8 t, uint8 t, uint8 t);
int main(int argc, char **argv) {
                                          Custom op declaration
  // First we'll declare some Vars to use below.
  Var x("x"), y("y"), c("c");
  // Take a color 8-bit input
  Image<uint8 t> input = load<uint8 t>("rgb.png");
  // Blur it horizontally:
  Func blur x("blur x");
  // Blur it vertically:
  Func blur y("blur y");
```



Case 2: Bilateral grid

- Edge preserving blur
- Starting point:
 - minimalistic scalar integer + float TTA processor
- Accelerated with custom ops:
 - Blur
 - 3D linear interpolation



Case 2: Bilateral grid

```
// Blur the grid using a five-tap filter
Func blurx("blurx"), blury("blury"), blurz("blurz");
```

```
blurz(x, y, z, c) = (histogram(x, y, z-2, c) +
                     histogram(x, y, z-1, c)*4 +
                     histogram(x, y, z , c)*6 +
                                                       Potential custom op
                     histogram(x, y, z+1, c)*4 +
                     histogram(x, y, z+2, c));
blurx(x, y, z, c) = (blurz(x-2, y, z, c) +
                     blurz(x-1, v, z, c)*4 +
                     blurz(x , y, z, c)*6 +
                     blurz(x+1, y, z, c)*4 +
                     blurz(x+2, y, z, c));
blury(x, y, z, c) = (blurx(x, y-2, z, c) +
                     blurx(x, y-1, z, c)*4 +
                     blurx(x, y , z, c)*6 +
                     blurx(x, y+1, z, c)*4 +
                     blurx(x, y+2, z, c));
```

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Case 2: Bilateral grid

```
HalideExtern 5(float, tce wavg5f, float, float, float, float, float);
int main(int argc, char **argv) {
    // Blur the grid using a five-tap filter
    Func blurx("blurx"), blury("blury"), blurz("blurz");
    blurz(x, y, z, c) = tce wavg5f(histogram(x, y, z-2, c),
                                    histogram(x, y, z-1, c),
                                    histogram(x, y, z , c),
                                    histogram(x, y, z+1, c),
                                    histogram(x, y, z+2, c));
    blurx(x, y, z, c) = tce wavg5f(blurz(x-2, y, z, c))
                                    blurz(x-1, v, z, c),
                                    blurz(x , y, z, c),
                                    blurz(x+1, v, z, c),
                                    blurz(x+2, y, z, c));
    blury(x, y, z, c) = tce_wavg5f(blurx(x, y-2, z, c))
                                    blurx(x, y-1, z, c),
                                    blurx(x, y , z, c),
                                    blurx(x, y+1, z, c),
                                    blurx(x, y+2, z, c));
```

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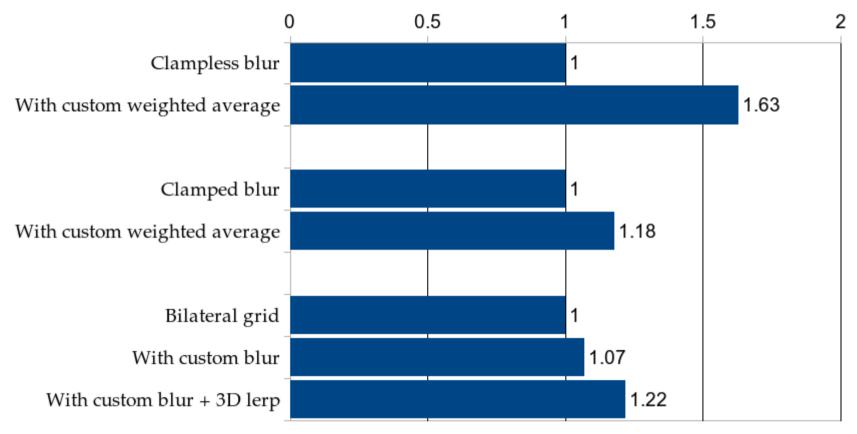
Linear interpolation

Original 3D linear interpolation implemented using Halide builtin function "lerp" (1D linear interpolation)

7 calls to lerp

Replaced with 3D lerp custom operation:

Results



speedup

- Clamped blur: address computations add overhead
- Bilateral grid: histogram computations dominate

Conclusion

- We described a tool flow from high abstraction level to customized processors
- Custom operations for increasing
 performance and/or energy efficiency
- Custom operations can be used directly from Halide descriptions
- Tool flow supports multicore custom processors; vector support in progress

