

# Why DSLs are Important to the DoE Exascale Mission

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# Outline

**Problem of (exa)Scaling High Performance Programs**

**Broader Impact of DSLs: The Halide Story**

**DoE Application Need for DSLs – Anshu Dubey**

# Evolving Code With Machines

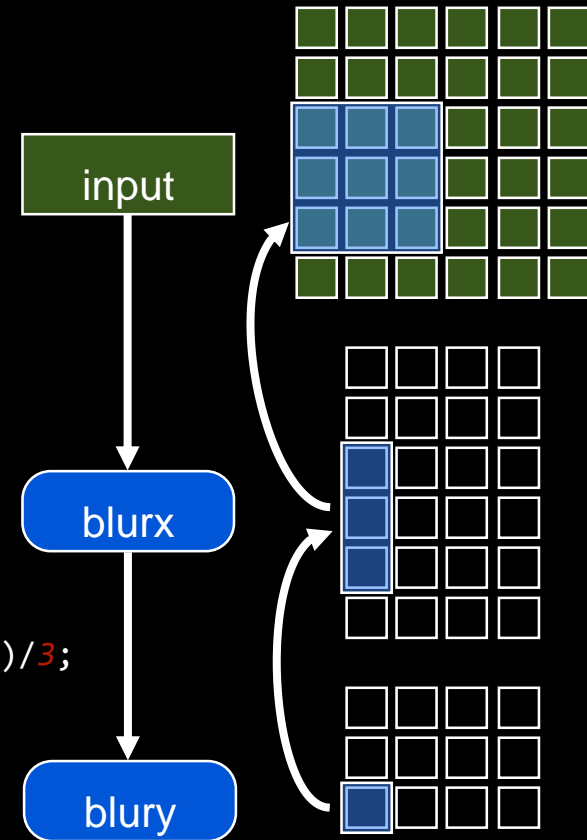
- Recent history: HPC machines evolving at incredible rate
- New topologies, new architectures, accelerators, GPUs, new interconnects, increased hierarchies, etc.
- Traditional approach means rewriting code to perform optimally on each new platform

# A simple example: 3x3 image blur

```
void box_filter_3x3(const Image &in, Image &blury) {  
    blurx(in.width(), in.height()); // allocate blurx array
```

```
    for (int y = 0; y < in.height(); y++)  
        for (int x = 0; x < in.width(); x++)  
            blurx(x, y) = (in(x-1, y) + in(x, y) + in(x+1, y))/3;
```

```
    for (int y = 0; y < in.height(); y++)  
        for (int x = 0; x < in.height(); x++)  
            blury(x, y) = (blurx(x, y-1) + blurx(x, y) + blurx(x, y+1))/3;
```



# Hand-optimized C++

**11x  
faster**  
(quad core x86)

```
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) {
        __m128i a, b, c, sum, avg;
        __m128i blurx[(256/8)*(32+2)]; // allocate tile blurx array
        for (int xTile = 0; xTile < in.width(); xTile += 256) {
            __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((__m128i*)(inPtr-1));
                    b = _mm_loadu_si128((__m128i*)(inPtr+1));
                    c = _mm_load_si128((__m128i*)(inPtr));
                    sum = _mm_add_epi16(_mm_add_epi16(a, b), c);
                    avg = _mm_mulhi_epi16(sum, one_third);
                    _mm_store_si128(blurxPtr++, avg);
                    inPtr += 8;
                }
            }
            blurxPtr = blurx;
            for (int y = 0; y < 32; y++) {
                __m128i *outPtr = (__m128i *)(&(blury[yTile+y][xTile]));
                for (int x = 0; x < 256; x += 8) {
                    a = _mm_load_si128(blurxPtr+(2*256)/8);
                    b = _mm_load_si128(blurxPtr+256/8);
                    c = _mm_load_si128(blurxPtr++);
                    sum = _mm_add_epi16(_mm_add_epi16(a, b), c);
                    avg = _mm_mulhi_epi16(sum, one_third);
                    _mm_store_si128(outPtr++, avg);
                }
            }
        }
    }
}
```

Tiled, fused

Vectorized

Multithreaded

Redundant  
computation

*Near roof-line  
optimum*

# (Re)organizing computation is hard

Optimizing parallelism, locality  
requires **transforming program &  
data structure.**

**What transformations are *legal*?**

**What transformations are  
*beneficial*?**

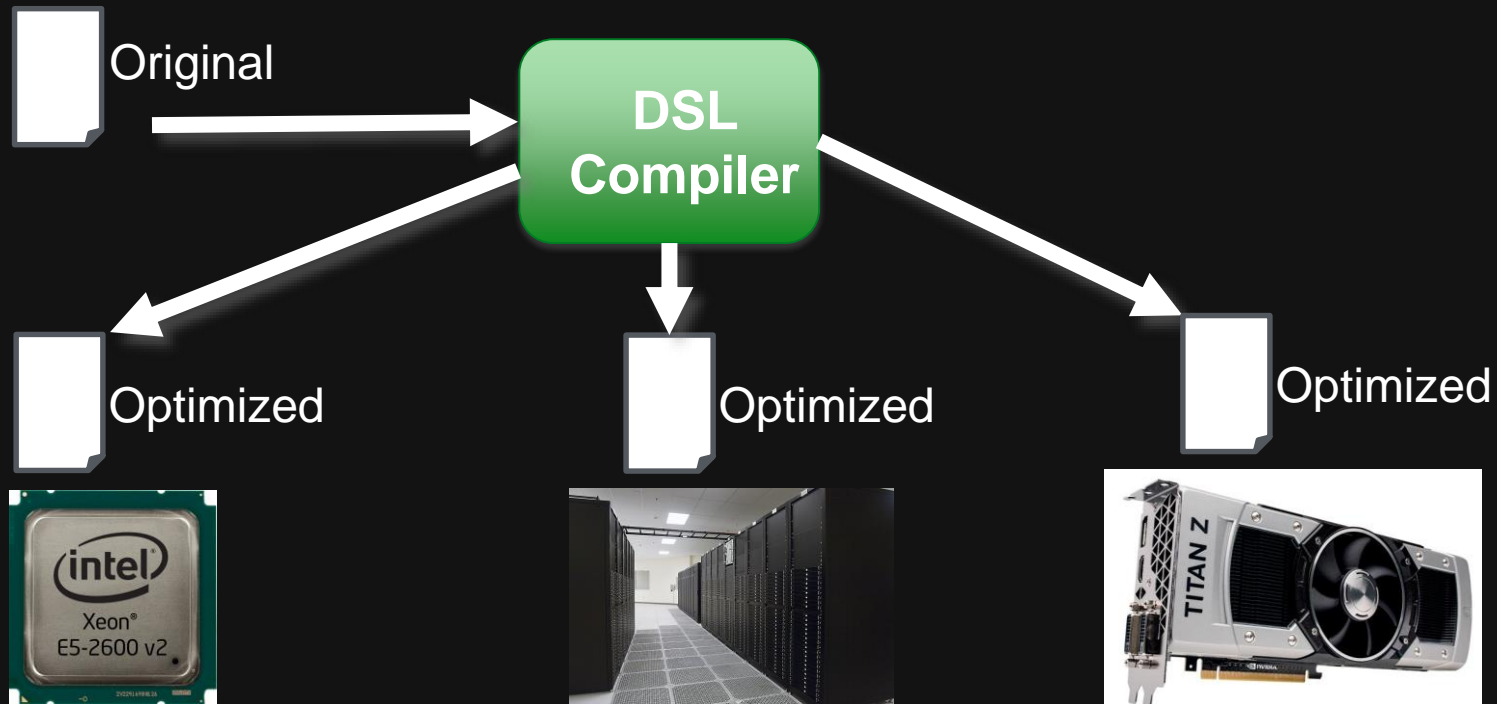
*Libraries don't solve this:*

**BLAS, IPP, MKL, OpenCV, MATLAB**

optimized kernels compose into inefficient pipelines (no  
fusion)

# Domain-Specific Languages: The Last Rewrite

- Instead of rewriting every application on every new platform, express as a set of domain-specific code
- Separate *algorithm* from optimizations



# Halide: Simpler, Faster, Scalable

**Reference: 300 lines C++**

with Jonathan Ragan-Kelley, Connelly Barnes,  
Andrew Adams, Sylvain Paris, Frédo Durand

**Adobe: 1500 lines**

*3 months of work*

*10x faster (vs. reference)*

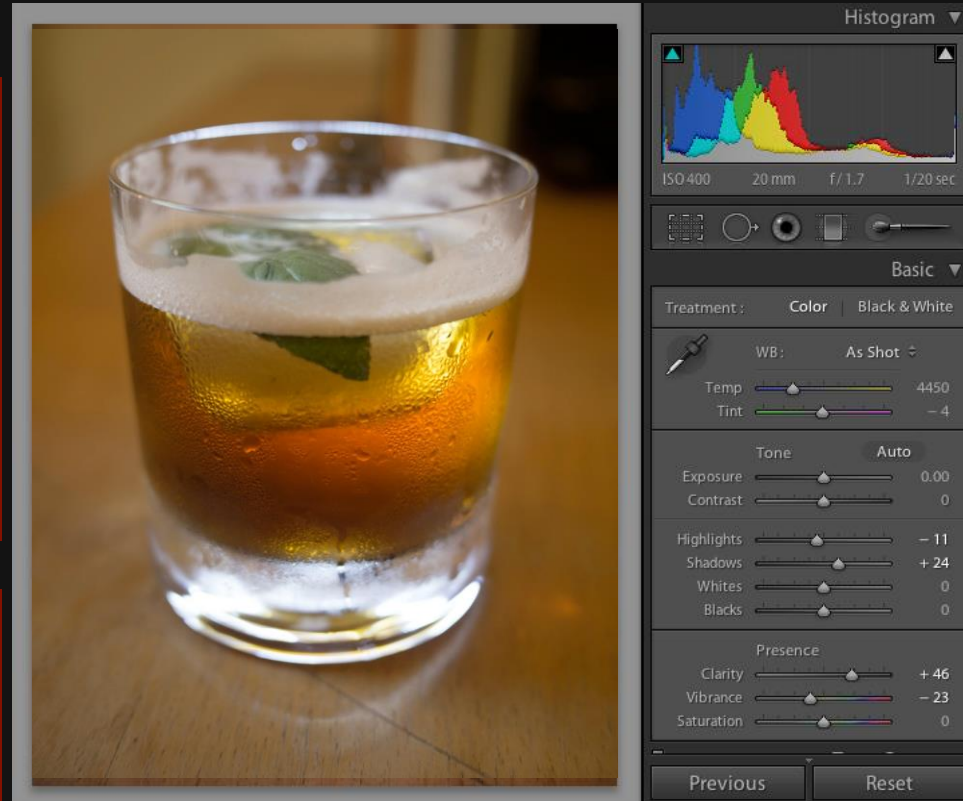
**Halide: 60 lines**

*1 intern-day*

**20x faster (vs. reference)**

**2x faster (vs. Adobe)**

**GPU: 90x faster**  
(vs. reference)



# Halide's answer:

*Decouple* algorithm from  
schedule

Algorithm: *what* is computed

Schedule: *where* and *when* it's computed

# Decoupling Algorithm from Schedule

## Optimized C++

```
void box_filter_3x3(const Image &in, Image &blurry) {
    __m128i one_third = _mm_set1_epi16(21846);
    #pragma omp parallel for
    for (int yTile = 0; yTile < in.height(); yTile += 32) {
        __m128i a, b, c, sum, avg;
        __m128i blurx[(256/8)*(32+2)]; // allocate tile blurx array
        for (int xTile = 0; xTile < in.width(); xTile += 256) {
            __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((__m128i*)(inPtr-1));
                    b = _mm_loadu_si128((__m128i*)(inPtr+1));
                    c = _mm_load_si128((__m128i*)(inPtr));
                    sum = _mm_add_epi16(_mm_add_epi16(a, b), c);
                    avg = _mm_mulhi_epi16(sum, one_third);
                    _mm_store_si128(blurxPtr++, avg);
                    inPtr += 8;
                }
            }
            blurxPtr = blurx;
            for (int y = 0; y < 32; y++) {
                __m128i *outPtr = (__m128i *)(&(blurry[yTile+y][xTile]));
                for (int x = 0; x < 256; x += 8) {
                    a = _mm_load_si128(blurxPtr+(2*256)/8);
                    b = _mm_load_si128(blurxPtr+256/8);
                    c = _mm_load_si128(blurxPtr++);
                    sum = _mm_add_epi16(_mm_add_epi16(a, b), c);
                    avg = _mm_mulhi_epi16(sum, one_third);
                    _mm_store_si128(outPtr++, avg);
                }
            }
        }
    }
}
```

## Halide I: Algorithm

```
void box_filter_3x3(const Image &in, Image &blurry) {
    Image blurx(in.width(), in.height()); // allocate blurx array

    for (int y = 0; y < in.height(); y++)
        for (int x = 0; x < in.width(); x++)
            blurx(x, y) = (in(x-1, y) + in(x, y) + in(x+1, y))/3;

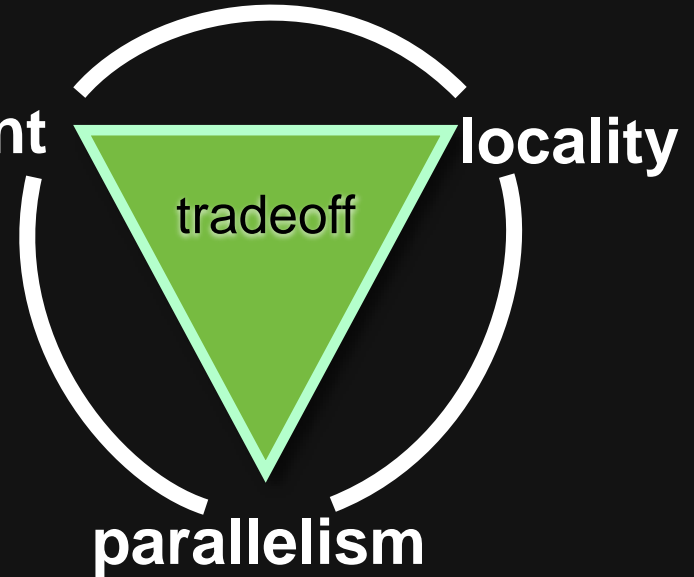
    for (int y = 0; y < in.height(); y++)
        for (int x = 0; x < in.width(); x++)
            blurry(x, y) = (blurx(x, y-1) + blurx(x, y) + blurx(x, y+1))/3;
}
```

# Halide II: Schedule

## Single Program, Multiple Optimization Paths

```
void box_filter_3x3(const Image &in, Image &blur) {  
    Image blurx(in.width(), in.height()); // allocate blurx array  
  
    for (int y = 0; y < in.height(); y++)  
        for (int x = 0; x < in.width(); x++)  
            blurx(x, y) = (in(x-1, y) + in(x, y) + in(x+1, y))/3;  
  
    for (int y = 0; y < in.height(); y++)  
        for (int x = 0; x < in.width(); x++)  
            blur(x, y) = (blurx(x, y-1) + blurx(x, y) + blurx(x,  
y+1))/3;  
}
```

redundant  
work



Once the algorithm is provided, can find the optimal schedule

Complex tradeoff space, no obvious winner

Each different tradeoff will lead to a “total rewrite of the program”

Best schedule depends....on architecture, rest of the program, inputs etc. etc.

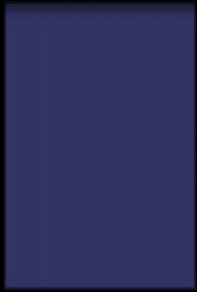
Can search the tradeoff space manually

ultra-fast Hypothesis-ScheduleGen-CodeGen-Evaluation cycle

Or can search the tradeoff space using autotuning

# Single Algorithm multiple schedules

in      blurx      blurry      in      blurx      blurry      in      blurx      blurry



redundant  
work



parallelism

locality

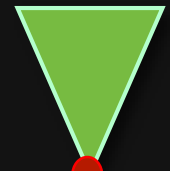
redundant  
work



parallelism

locality

redundant  
work



parallelism

locality

in      blurx      blurry      in      blurx      blurry      in      blurx      blurry



redundant  
work



parallelism

locality

redundant  
work



parallelism

locality

redundant  
work

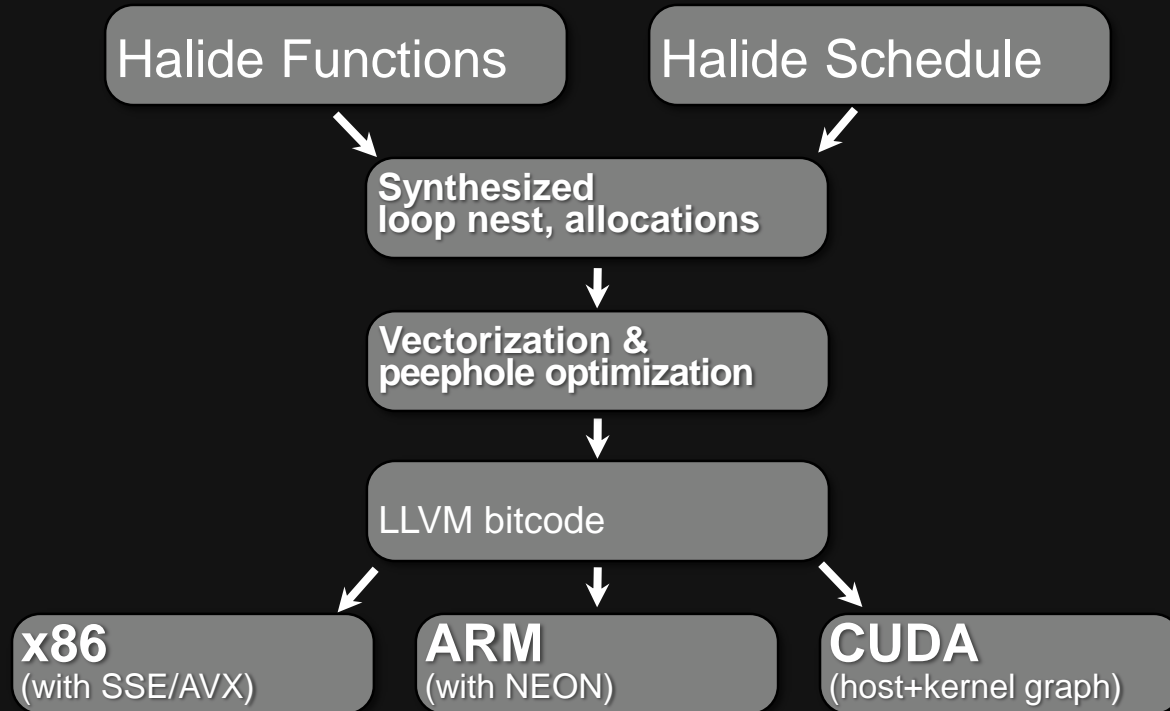


parallelism

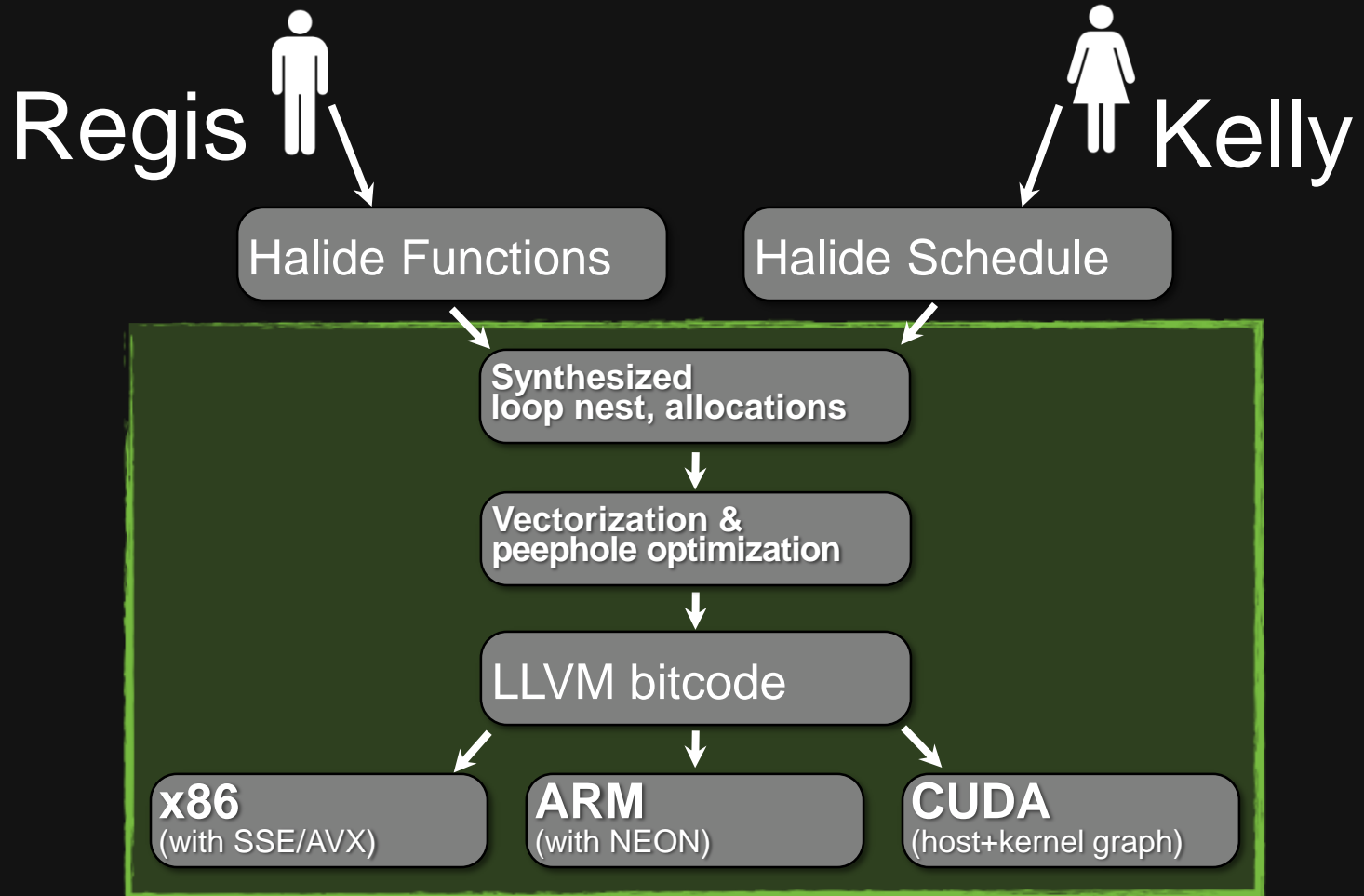
locality

D-TEC

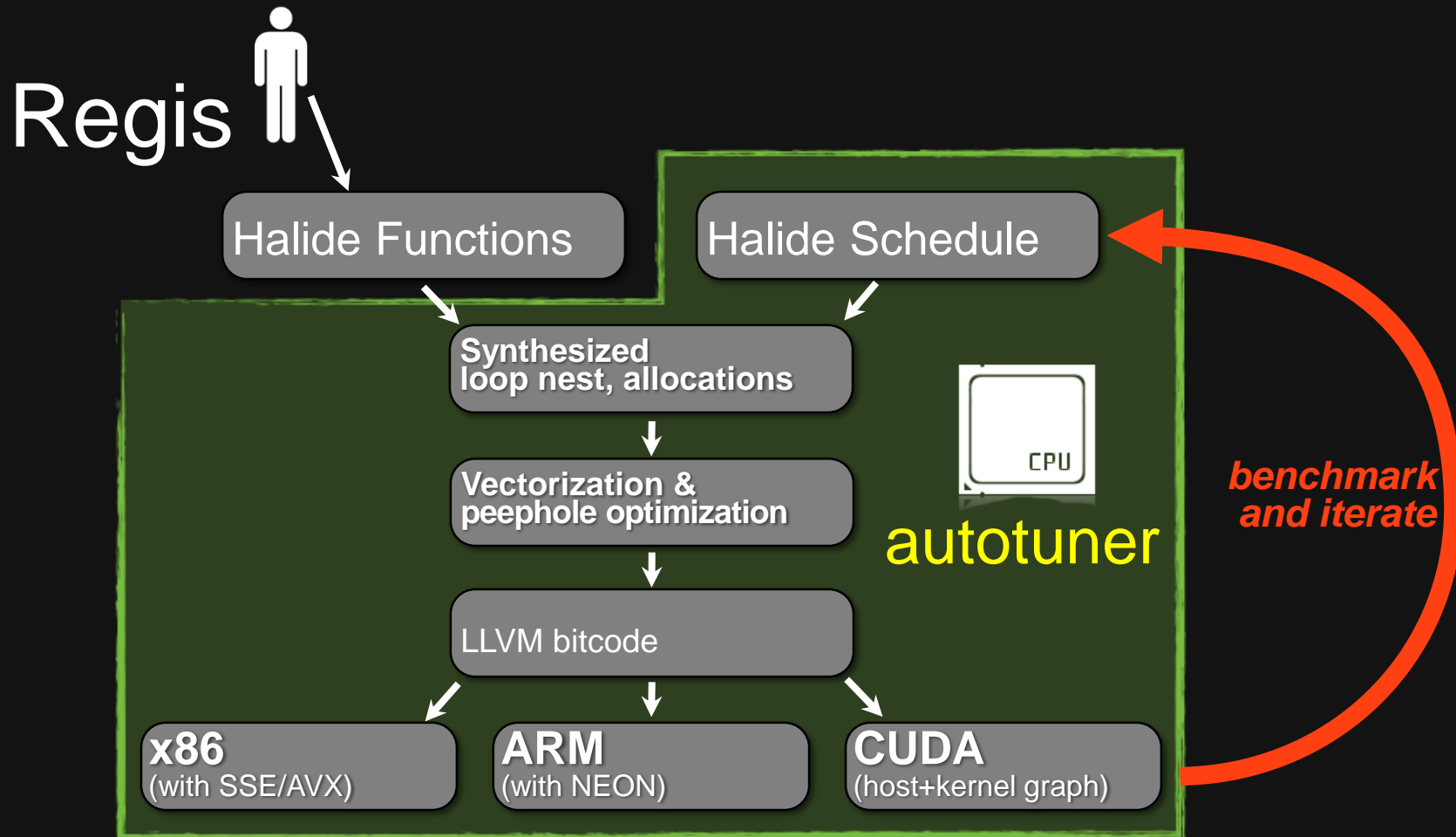
# The Halide Compiler



# The Halide Compiler



# The Halide Compiler



Uses OpenTuner framework  
(at the technology marketplace)

x86	Speedup	Factor shorter
Blur	1.2 ×	18 ×
Bilateral Grid	4.4 ×	4 ×
Camera pipeline	3.4 ×	2 ×
“Healing brush”	1.7 ×	7 ×
Local Laplacian	1.7 ×	5 ×

GPU	Speedup	Factor shorter
Bilateral Grid	2.3 ×	11 ×
“Healing brush”	5.9*	7*
Local Laplacian	9*	7*

ARM	Speedup	Factor shorter
Camera pipeline	1.1 ×	3 ×

**Autotuning time: 2 hrs to 2 days**  
 (single node) **85% within < 24 hrs**

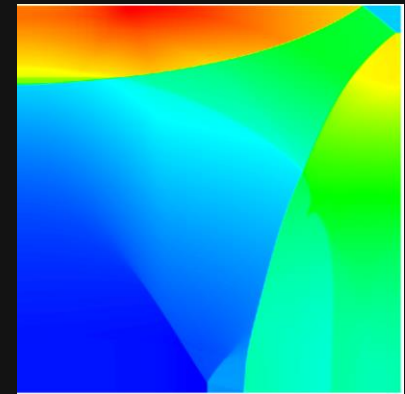
# Mini app 1: Cloverleaf

Mantevo benchmark implementing Lagrangian-Eulerian hydrodynamics in 2D

>40% of runtime spent in `advec_mom` computational kernel (~255 LoC)

`advec_mom` ported to Halide and run on a 2x12 core Ivy Bridge Xeon

Ported to Halide by Shoaib Kamil



# Mini app 1: Cloverleaf

## Original OpenMP (partial)

```
!$OMP DO
DO k=y_min-2,y_max+2
DO j=x_min-2,x_max+2
post_vol(j,k)= volume(j,k)+vol_flux_y(j ,k+1)-vol_flux_y(j,k)
pre_vol(j,k)=post_vol(j,k)+vol_flux_x(j+1,k )-vol_flux_x(j,k)
ENDDO
!$OMP DO
DO k=y_min,y_max+1
DO j=x_min-2,x_max+2
! Find staggered mesh mass fluxes, nodal masses and volumes.
node_flux(j,k)=0.25_8*(mass_flux_x(j,k-1 )+mass_flux_x(j ,k) &
+mass_flux_x(j+1,k-1)+mass_flux_x(j+1,k)) ! Mass Flux
ENDDO
!$OMP END DO
!$OMP DO
DO k=y_min,y_max+1
DO j=x_min-1,x_max+2
! Staggered cell mass post advection
node_mass_post(j,k)=0.25_8*(density1(j ,k-1)*post_vol(j ,k-1)
+density1(j ,k )*post_vol(j ,k )
+density1(j-1,k-1)*post_vol(j-1,k-1)
+density1(j-1,k )*post_vol(j-1,k ))
ENDDO
!$OMP END DO
!$OMP DO
DO k=y_min,y_max+1
DO j=x_min-1,x_max+2
! Staggered cell mass pre advection
node_mass_pre(j,k)=node_mass_post(j,k)-node_flux(j-1,k)+node_flux(j,k)
ENDDO
!$OMP END DO
!$OMP PRIVATE(upwind,downwind,donor,dif,sigma,width,limiter,vdiffw,vdiffdw,auw,adw,wind)
DO k=y_min,y_max+1
DO j=x_min-1,x_max+1
IF (node_flux(j,k).LT.0.0)THEN
upwind=j+2
donor=j+1
downwind=j
dif=donor
ELSE
upwind=j-1
donor=j
downwind=j+1
dif=upwind
ENDIF
sigma=ABS(node_flux(j,k))/(node_mass_pre(donor,k))
width=celldx(j)
vdiffw=vell(donor,k)-vell(upwind,k)
vdiffdw=vell(downwind,k)-vell(donor,k)
limiter=0.0
IF (vdiffw*vdiffdw.GT.0.0)THEN
auw=ABS(vdiffw)
adw=ABS(vdiffdw)
wind=1.0_8
IF (vdiffdw.LE.0.0) wind=-1.0_8
limiter=wind*MIN(width*((2.0_8-sigma)*adw/width+(1.0_8+sigma)*auw/celldx(dif)))/6.0_8,auw,adw)
ENDIF
advec_vel(j,k)=vell(donor,k)+(1.0-sigma)*limiter
mom_flux(j,k)=advec_vel(j,k)*node_flux(j,k)
ENDDO
!$OMP END DO
!$OMP DO
DO k=y_min,y_max+1
DO j=x_min,x_max+1
vell (j,k)=(vell (j,k)*node_mass_pre(j,k)+mom_flux(j-1,k)-mom_flux(j,k))/node_mass_post(j,k)
ENDDO
!$OMP END DO
```

## Halide (program only)

```
Expr e_post_vol = volume(j,k) + vol_flux_y(j,k+1) - vol_flux_y(j,k);
f_post_vol(j,k) = e_post_vol;

Expr e_pre_vol = f_post_vol(j,k) + vol_flux_x(j+1,k) - vol_flux_x(j,k);
f_pre_vol(j,k) = e_pre_vol;

Expr e_node_flux = 0.25f * (mass_flux_x(j,k-1)
+ mass_flux_x(j,k)
+ mass_flux_x(j+1,k-1)
+ mass_flux_x(j+1,k));
f_node_flux(j,k) = e_node_flux;

Expr e_node_mass_post = 0.25f * (density1(j,k-1) * f_post_vol(j,k-1)
+ density1(j,k) * f_post_vol(j,k)
+ density1(j-1,k-1) * f_post_vol(j-1,k-1)
+ density1(j-1,k) * f_post_vol(j-1,k));
f_node_mass_post(j,k) = e_node_mass_post;

Expr e_node_mass_pre = f_node_mass_post(j,k) - f_node_flux(j-1,k) + f_node_flux(j,k);
f_node_mass_pre(j,k) = e_node_mass_pre;

Expr upwind = select(f_node_flux(j,k) < 0.0f, j+2, j-1);
Expr donor = select(f_node_flux(j,k) < 0.0f, j+1, j);
Expr downwind = select(f_node_flux(j,k) < 0.0f, j, j-1);
Expr dif = select(f_node_flux(j,k) < 0.0f, donor, upwind);

Expr sigma = abs(f_node_flux(j,k)) / f_node_mass_pre(donor, k);
Expr width = celldx(j);
Expr vdiffw = vell(donor,k) - vell(upwind,k);
Expr vdiffdw = vell(downwind,k) - vell(donor,k);

Expr auw = abs(vdiffw);
Expr adw = abs(vdiffdw);
Expr wind = select(vdiffdw <= 0.0f, -1.0f, 1.0f);

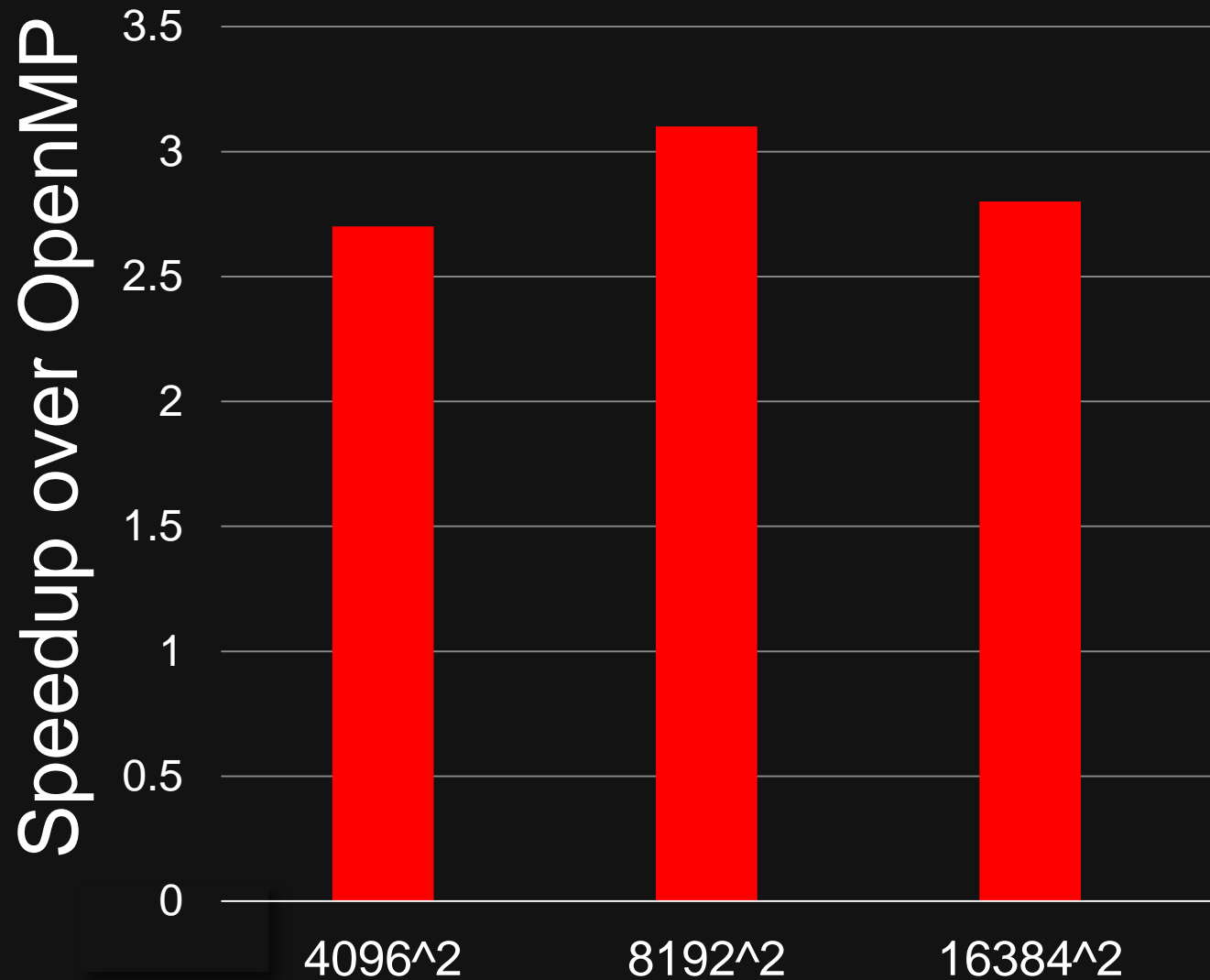
Expr limiter = select(vdiffw*vdiffdw > 0.0f, wind*min(width*((2.0f-sigma)*
adw/width+(1.0f+sigma)*auw/celldx(dif))/6.0f,
min(auw,adw)),
cast(Float(64), 0.0f));

Expr e_advec_vel = vell(donor,k) + (1.0f-sigma)*limiter;
f_advec_vel(j,k) = e_advec_vel;

Expr e_mom_flux = f_advec_vel(j,k) * f_node_flux(j,k);
f_mom_flux(j,k) = e_mom_flux;

Expr e_vell = (vell(j,k) * f_node_mass_pre(j,k)
+ f_mom_flux(j-1,k)
- f_mom_flux(j,k)) / f_node_mass_post(j,k);
```

# Mini app 1: Cloverleaf



# Mini app 2: CNS solver

**Compressible Navier Stokes equation for Constant Viscosity and Thermal Conductivity.**

## Halide Port

- Ported two kernal to Halide which consumes >50% of the run time (diffterm and hyp term).
- Used an auto-tuner to obtain high performance schedules for a SMP CPU
- Used a hand-generated schedule to get GPU results
- Halide uses loop fusion in diffterm VS. original fortran code to outperform in all kernal sizes that were tested

Ported to Halide by Charith Mendis

# Mini app 2: CNS solver

Fortran code

A screenshot of Fortran code for the CNS solver mini app. The code is written in a standard Fortran style with many lines of text, including variable declarations, loops, and conditional statements. It is color-coded with blue for keywords and black for identifiers and literals.

Halide code

A screenshot of Halide code for the CNS solver mini app. The code is written in a Halide-specific syntax, which is more concise than Fortran. It uses Halide's expression-based syntax for computations and loops, with color-coding similar to Fortran.

C code

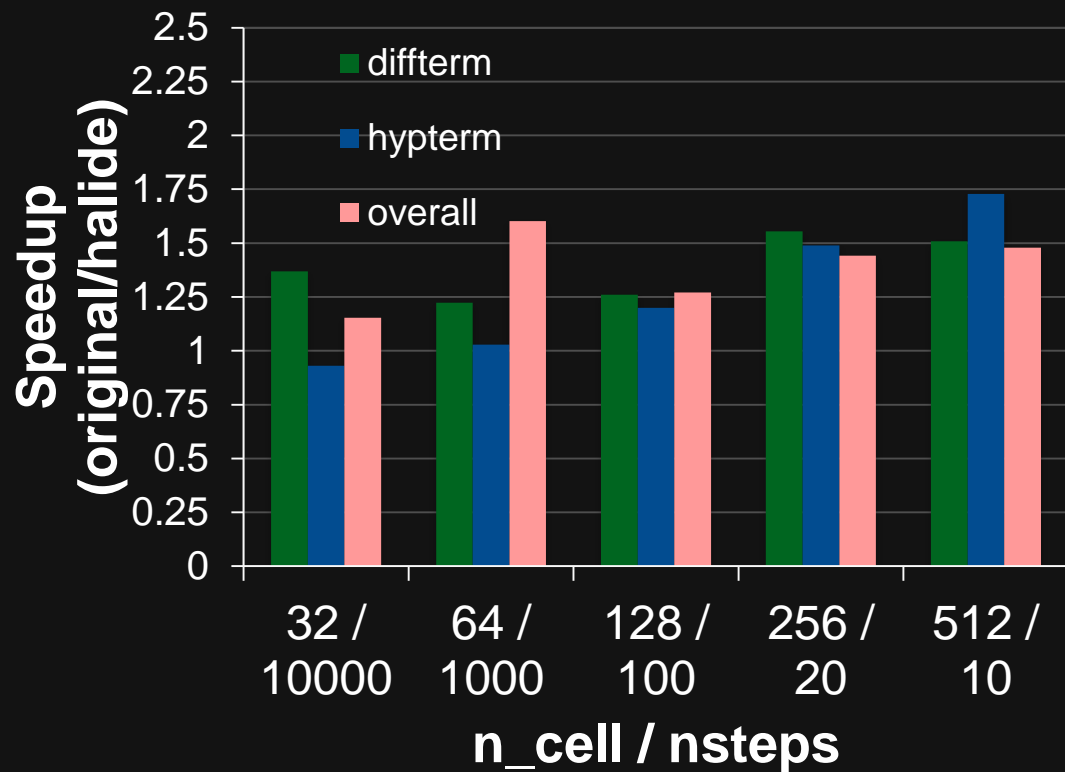
A screenshot of C code for the CNS solver mini app. The code is written in a standard C style with many lines of text, including variable declarations, loops, and conditional statements. It is color-coded with blue for keywords and black for identifiers and literals.

Halide advantages

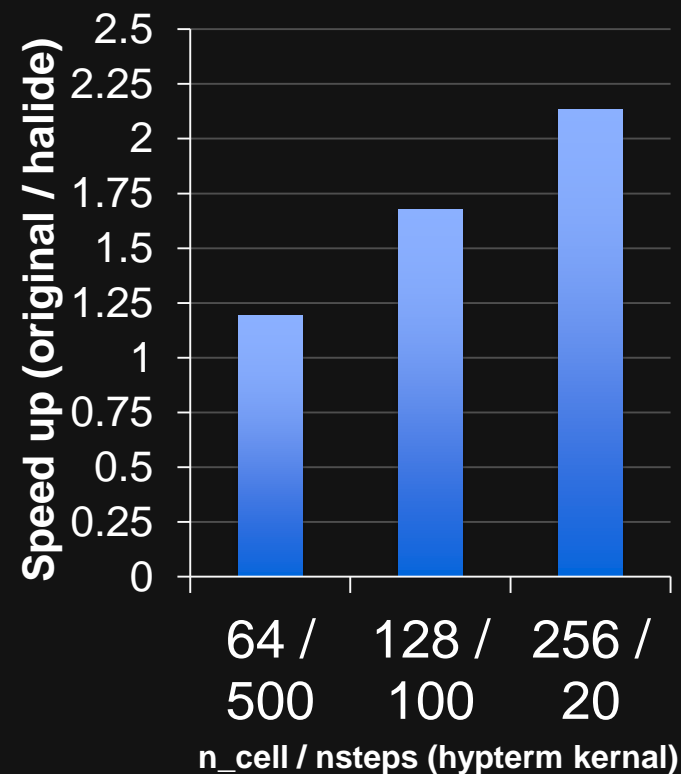
1. Can be directly ported without syntax changes from Fortran to Halide
2. Ability to try out various schedules without obscuring the algorithm
3. Lines of Code approximately equal to Fortran code and is much less than the equivalent C code

# Mini app 2: CNS solver

## CPU

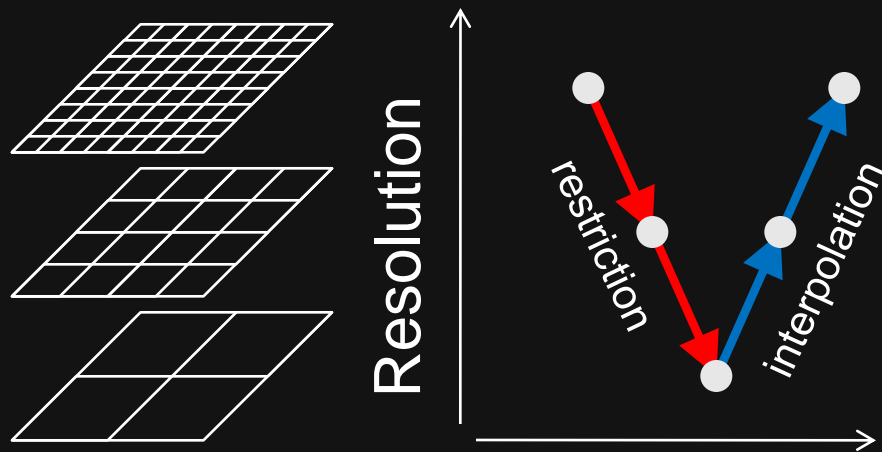


## GPU



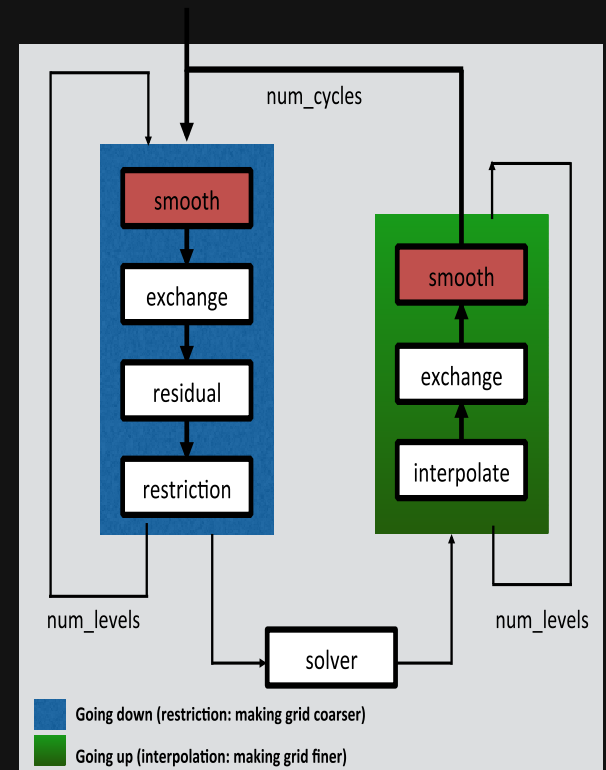
# Miniapp 3: HPGMG

HPGMG is a compact benchmark designed to proxy the geometric MG solves found in applications built from AMR MG frameworks like CHOMBO or BoxLib.



Started with Sam's hand optimized code with fusion as our ORIGINAL

Ported to Halide by Rishi Khan



# Mini app 3: HPGMG

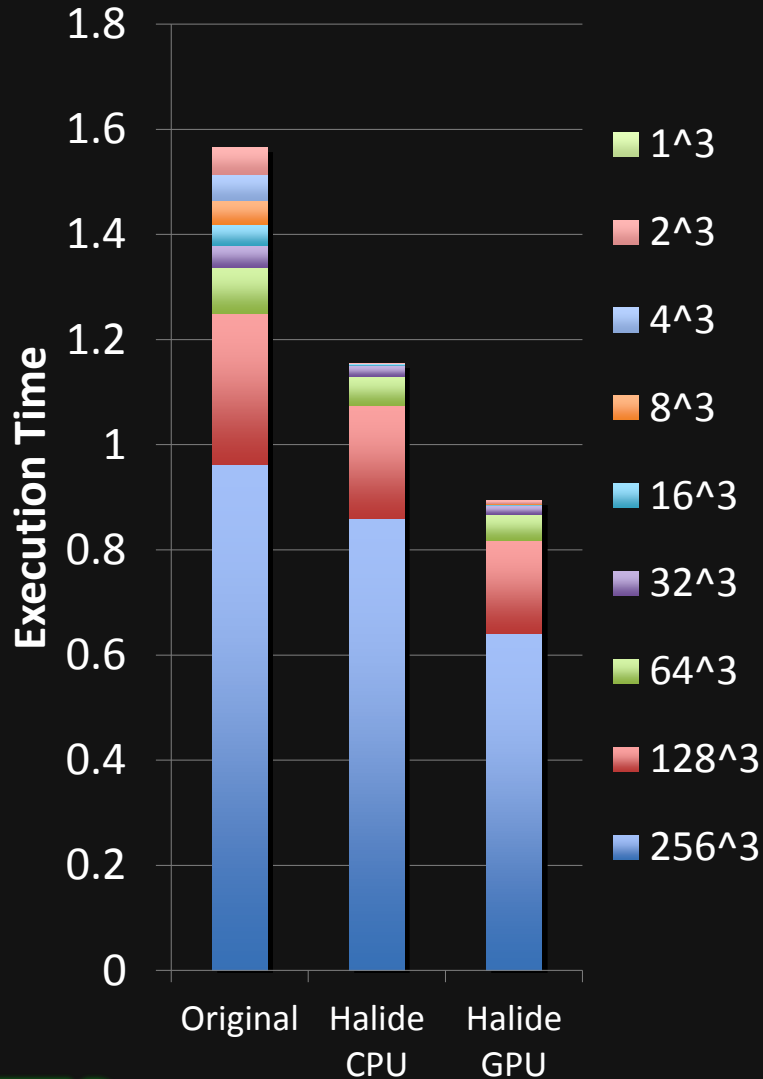
Original Hand optimized

Halide program

```
#define OMP_THREAD_WITHIN_A_BOX(threads_per_team) \  
    if(threads_per_team>1) num_threads(threads_per_team) collapse(2)  
int i,j,k;  
#pragma omp parallel for private(k,j,i) OMP_THREAD_WITHIN_A_BOX(level->threads_per_box)  
for(k=0-ghostsToOperateOn;k<dim+ghostsToOperateOn;k++) {  
    for(j=0-ghostsToOperateOn;j<dim+ghostsToOperateOn;j++) {  
        for(i=0-ghostsToOperateOn;i<dim+ghostsToOperateOn;i++) {  
            int ijk = i + j*jStride + k*kStride;  
            double Ax_n = a*alpha[ijk]*x_n[ijk] - b*h2inv*(  
                beta_i[ijk] *(valid[ijk-1] *(x_n[ijk] + x_n[ijk-1]) - 2.0*x_n[ijk])  
                + beta_j[ijk] *(valid[ijk-jStride] *(x_n[ijk] + x_n[ijk-jStride]) - 2.0*x_n[ijk])  
                + beta_k[ijk] *(valid[ijk-kStride] *(x_n[ijk] + x_n[ijk-kStride]) - 2.0*x_n[ijk])  
                + beta_i[ijk+1] *(valid[ijk+1] *(x_n[ijk] + x_n[ijk+1]) - 2.0*x_n[ijk])  
                + beta_j[ijk+jStride] *(valid[ijk+jStride] *(x_n[ijk] + x_n[ijk+jStride]) - 2.0*x_n[ijk])  
                + beta_k[ijk+kStride] *(valid[ijk+kStride] *(x_n[ijk] + x_n[ijk+kStride]) -  
                2.0*x_n[ijk]));  
            double lambda = 1.0 / (a*alpha[ijk] - b*h2inv*(  
                beta_i[ijk] *(valid[ijk-1] - 2.0)  
                + beta_j[ijk] *(valid[ijk-jStride] - 2.0)  
                + beta_k[ijk] *(valid[ijk-kStride] - 2.0)  
                + beta_i[ijk+1] *(valid[ijk+1] - 2.0)  
                + beta_j[ijk+jStride] *(valid[ijk+jStride] - 2.0)  
                + beta_k[ijk+kStride] *(valid[ijk+kStride] - 2.0)  
                ));  
            x_npl[ijk] = x_n[ijk] + c1*(x_n[ijk]-x_nml[ijk]) + c2*lambda*(rhs[ijk]-Ax_n);  
        })  
    }  
}  
  
//... scheduling constraints in a different file  
level->concurrent_boxes = level->num_my_boxes;  
if(level->concurrent_boxes > omp_threads) level->concurrent_boxes = omp_threads;  
if(level->concurrent_boxes < 1) level->concurrent_boxes = 1;  
level->threads_per_box = omp_threads / level->concurrent_boxes;  
if(level->threads_per_box > level->box_dim*level->box_dim)  
    level->threads_per_box = level->box_dim*level->box_dim; // JK collapse  
if(level->threads_per_box > level->box_dim*level->box_dim*level->box_dim/64)  
    level->threads_per_box = level->box_dim*level->box_dim*level->box_dim/64;  
if(level->threads_per_box<1) level->threads_per_box = 1;
```

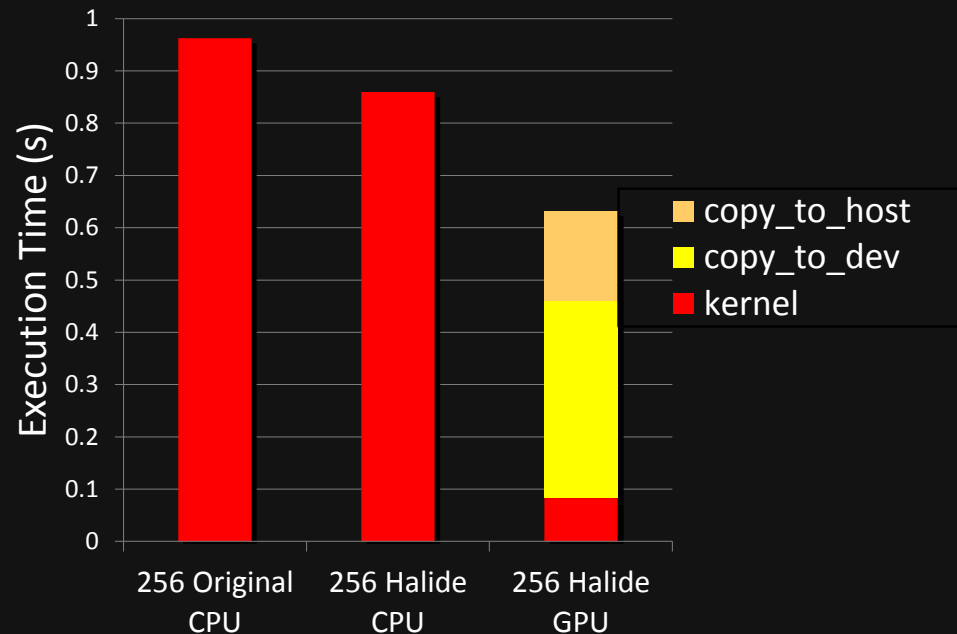
```
Func Ax_n("Ax_n"), lambda("lambda"), chebyshev("chebyshev");  
Var i("i"), j("j"), k("k");  
Ax_n(i,j,k) = a*alpha(i,j,k)*x_n(i,j,k) - b*h2inv*(  
    beta_i(i,j,k) *(valid(i-1,j,k)*(x_n(i,j,k) + x_n(i-1,j,k)) - 2.0f*x_n(i,j,k))  
    + beta_j(i,j,k) *(valid(i,j-1,k)*(x_n(i,j,k) + x_n(i,j-1,k)) - 2.0f*x_n(i,j,k))  
    + beta_k(i,j,k) *(valid(i,j,k-1)*(x_n(i,j,k) + x_n(i,j,k-1)) - 2.0f*x_n(i,j,k))  
    + beta_i(i+1,j,k)*(valid(i+1,j,k)*(x_n(i,j,k) + x_n(i+1,j,k)) - 2.0f*x_n(i,j,k))  
    + beta_j(i,j+1,k)*(valid(i,j+1,k)*(x_n(i,j,k) + x_n(i,j+1,k)) - 2.0f*x_n(i,j,k))  
    + beta_k(i,j,k+1)*(valid(i,j,k+1)*(x_n(i,j,k) + x_n(i,j,k+1)) - 2.0f*x_n(i,j,k)));  
lambda(i,j,k) = 1.0f / (a*alpha(i,j,k) - b*h2inv*(  
    beta_i(i,j,k) *(valid(i-1,j,k) - 2.0f)  
    + beta_j(i,j,k) *(valid(i,j-1,k) - 2.0f)  
    + beta_k(i,j,k) *(valid(i,j,k-1) - 2.0f)  
    + beta_i(i+1,j,k)*(valid(i+1,j,k) - 2.0f)  
    + beta_j(i,j+1,k)*(valid(i,j+1,k) - 2.0f)  
    + beta_k(i,j,k+1)*(valid(i,j,k+1) - 2.0f)));  
chebyshev(i,j,k) = x_n(i,j,k) + c1*(x_n(i,j,k)-x_nml(i,j,k))+  
    c2*lambda(i,j,k)*(rhs(i,j,k)-Ax_n(i,j,k));
```

# Mini app 3: HPGMG



## Speedups

Stencil Size	Halide CPU	Halide GPU
256^3	1.12	1.50
128^3	1.34	1.63
64^3	1.54	1.73
32^3	2.14	2.75
16^3	11.52	11.31
8^3	67.43	67.73
4^3	326.96	143.10
2^3	548.78	8.89



# Halide's Broader Impact

Google Glass camera pipeline

Google+ image pipeline



Exploring use in other Google projects

- 20+ engineers writing halide code & 2 experts writing the schedules

Adobe also exploring use in products

