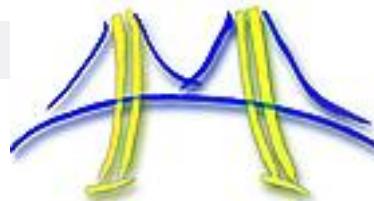
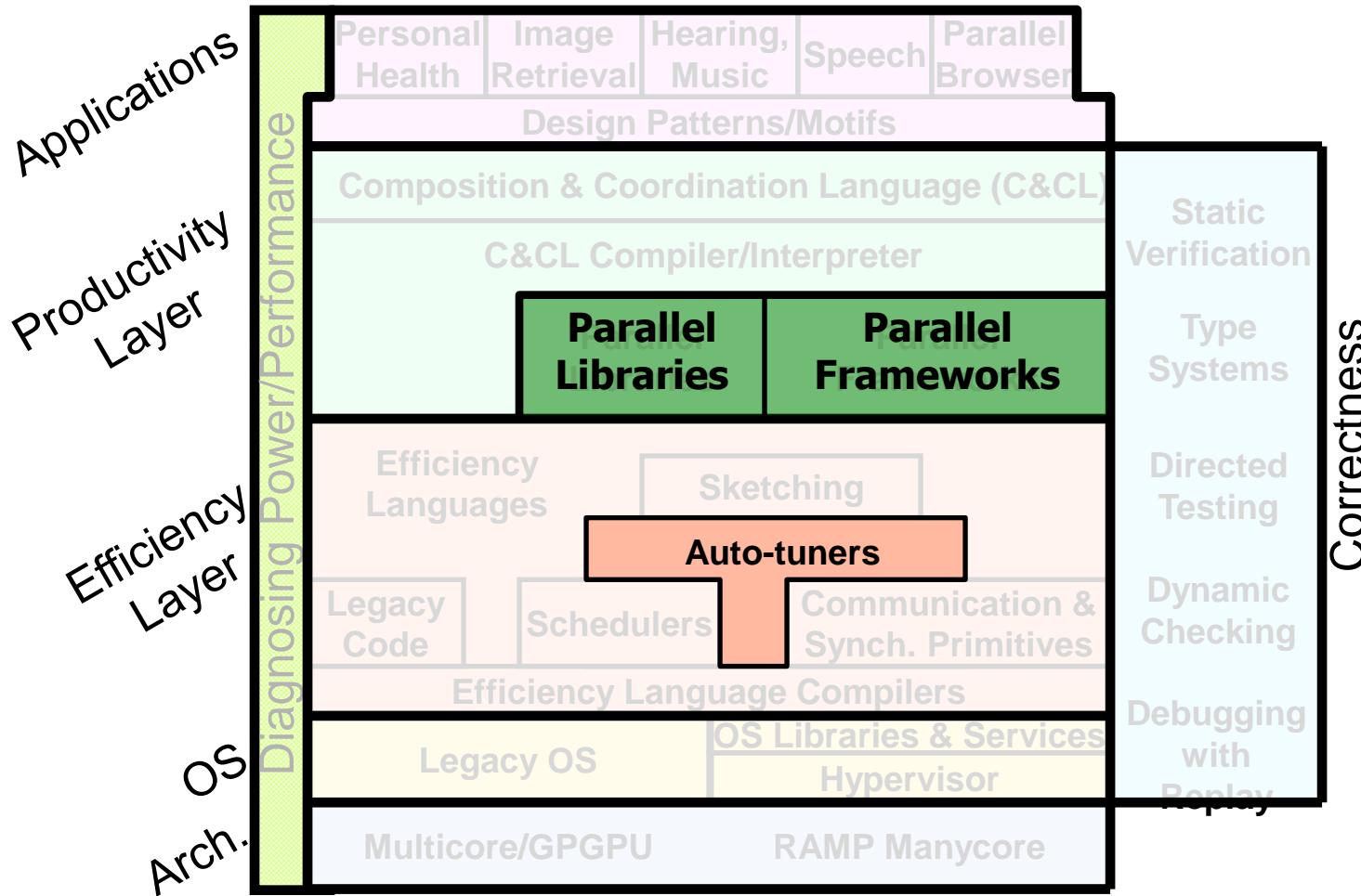


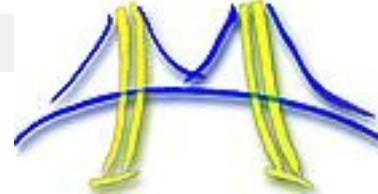
Code Generators for Stencil Auto-tuning

Shoaib Kamil with Cy Chan, John Shalf,
Sam Williams, Kaushik Datta, Katherine Yelick,
Jim Demmel, Leonid Oliker



Where this fits in Parlab





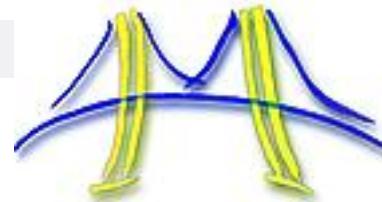
Making Auto-tuners Really Auto

■ Search

- Old: **Extremely user guided**, or arbitrary orderings of optimizations
- New: more intelligence (as in talk by K. Datta & A. Ganapathi)

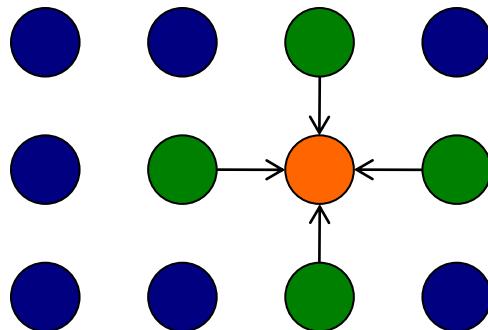
■ Code generation

- Old: Perl scripts (**one per kernel**) that are essentially “glorified printf” aka string substitution
- New: represent kernels abstractly (this talk), extend auto-tuning to motifs that are not well-represented as libraries

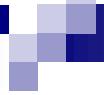


What is a Stencil Kernel?

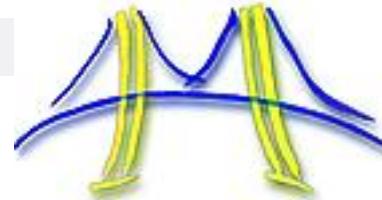
- Many computations on grids with regular structures can be represented as “sweeps” over the grid, where each point in a sweep is a arithmetic combination of the point’s neighbors
- “Kernel” is an instance of a stencil operator
- Want to make Auto-tuners for many (all?) motifs; start with stencils
 - Varied enough but still relatively simple



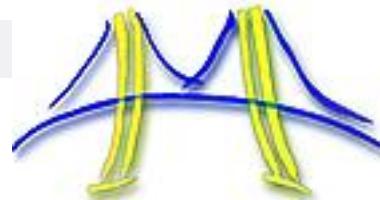
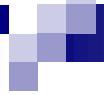
Stencil Example: 5 point stencil in 2D. Orange point is updated with a combination of the green points.



Goals

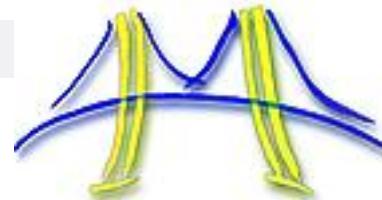


- How can we take previous work using a single stencil kernel and build an auto-tuning system like Atlas/PhiPAC/OSKI
 - Relatively user-friendly
 - Automated
 - Works across many stencil kernels
 - Target GPUs, Manycore, SMP
- Too much variability to write a single library that contains all possible kernels (a la OSKI)
- What can we learn about productively writing auto-tuners
- Proof-of-Concept



Overview of Stencil Auto-tuner

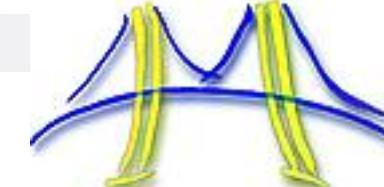
- Front end parses stencil kernel using Domain-Specific Language
- Transformation framework applies domain-specific optimizations
- Code generator outputs candidate versions
- Controlled by Search system ("outer loop")
- We do the dirty work of optimizing:
Performance Layer programmer can just run the Stencil Auto-tuner to get **performance portability**



DSL's for Stencils?

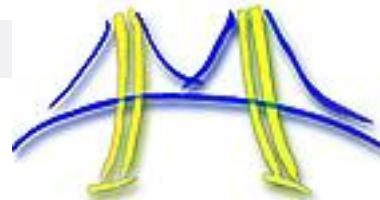
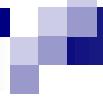
- Only need a few features:
 - Represent the operation at each point
 - E.g. "add left and right neighbors, multiply by 4"
 - Represent operations at boundaries
- Why not use existing code, say Fortran or Matlab + annotations?

```
do i=2,99,1
    do j=2,99,1
        wr_arry(i,j) = rd_arry(i,j) + rd_arry(i+1,j) -
                        rd_arry(i,j+1)
    enddo
enddo
```



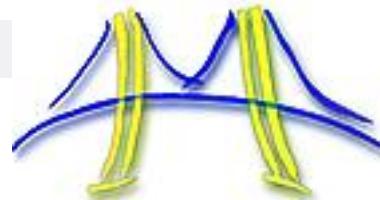
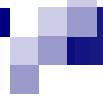
Domain-Specific Transformations

- Express optimizations as transformations to an abstract representation of the stencil
 - E.g. “Cache Blocking” is a combination of two loop transformations
- Advantage: can arbitrarily mix/combine optimizations
- Domain-specificity means minimal analysis needed
 - Only do transformations we’re sure are correct
 - Example: if we know stencil “footprint” can tell how/whether to cache block



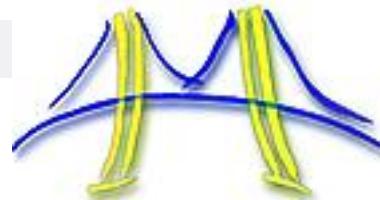
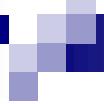
Transformations

- Proof-of-Concept represents stencil kernel as simplified AST
- AST -> AST transformations
- Alternative methods
 - Sketching (some investigation of this already)
 - Low-Level Virtual Machine (LLVM)
 - String substitution/rewriting
 - Probably not powerful enough
 - Others?



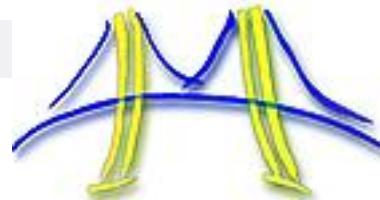
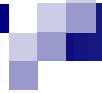
Code Generation

- Current backends do AST->code
 - Fortran (serial)
 - C (serial, several different array representations)
 - C + pthreads
 - CUDA
- Parallel backends represent parallelization as just another transformation
- In progress/Consideration
 - C + Cell-specific pthreads
 - OpenCL in progress
 - UPC? Fortress or X10?



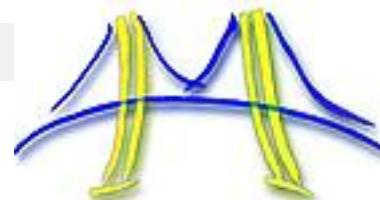
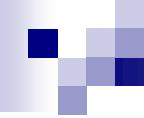
Search

- Search system goals
 - Should automatically extract annotated kernels
 - Then run candidate implementations
 - Gather “winning” implementations (one per kernel) into a library that can be called by the application
- Currently, the “outer loop” runs candidate implementations and uses fastest
- Semi-exhaustive search (e.g. powers of 2 cache blocking)
- More intelligent search needed?
 - Hill climbing? (a la CG)
 - Machine learning? (see talk by K. Datta & A. Ganapathi)



Status of Proof-of-Concept

- Few months of effort
- Implemented most transformations from K. Datta & S. Williams SC08 paper
- Many lessons learned
 - Writing code is important!
 - Testing with real-world kernels & applications
 - Use not only microbenchmarks like Stencil Probe
 - Green Flash's climate application has many stencils-> many tests
 - Implementation in Lisp ("the L-word!")
 - Higher order functions make composition easier
 - Simple tree representation as lists
 - Mostly-functional programming
 - Don't worry: Lisp -> C then linked with C libraries



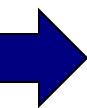
Auto-tuning Example

- Operators extracted from climate code before and after auto-tuning

```

do k=0,km,1
  do iprime=1,nside,1
    do i=2,im2nghost-1,1
      ia = i + ii(iprime)
      do j=2,jm2nghost-1,1
        ja = j + jj(iprime)
        buoyancy_gen(i,j,iprime,k)
        = -1.0*g*(theta(ia,ja,k)
                    - theta(i,j,k))
        / (theta00(k)*el(iprime))
      enddo
    enddo
  enddo
enddo

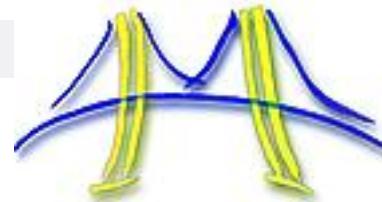
```



```

do G14906=0,km,4
do G14907=1,nside,6
do G14908=2,im2nghost - 1,25
do G14909=2,jm2nghost - 1,25
do k=G14906,G14906 + 1,1
do iprime=G14907,G14907 + 5,1
do i=G14908,G14908 + 24,1
ia = i + ii(iprime)
do j=G14909,G14909 + 24,1
ja = j + jj(iprime)
buoyancy_gen(i,j,iprime,k) = -1.0 * g * theta(ia,ja,k) - theta(i,j,k) /
theta00(k) * el(iprime)
enddo
enddo
enddo
enddo
enddo
enddo
enddo
do G14907=1,nside,6
do G14908=2,im2nghost - 1,25
do G14909=2,jm2nghost - 1,25
do k=G14906 + 2,G14906 + 3,1
do iprime=G14907,G14907 + 5,1
do i=G14908,G14908 + 24,1
ia = i + ii(iprime)
do j=G14909,G14909 + 24,1
ja = j + jj(iprime)
buoyancy_gen(i,j,iprime,k) = -1.0 * g * theta(ia,ja,k) - theta(i,j,k) /
theta00(k) * el(iprime)
enddo
enddo
enddo
enddo
enddo
enddo
enddo

```



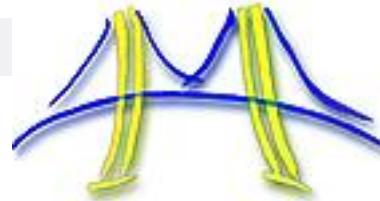
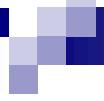
Examples

■ Heat

$$A_{out}(i,j,k) = A(i,j,k) + \text{factor} * (A(i+1,j,k) + A(i-1,j,k) + A(i,j+1,k) + A(i,j-1,k) + A(i,j,k+1) + A(i,j,k-1))$$

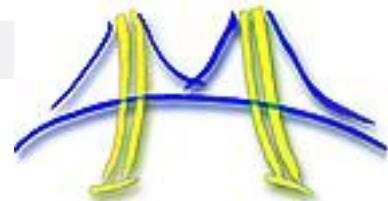
■ Divergence (from application)

$x_{out}(i,j) =$ &
(v_weights(1,2,1,i ,j) * xin(1,1,i ,j) &
- v_weights(1,1,1,i ,j) * xin(2,1,i ,j) &
+ v_weights(1,2,2,i ,j) * xin(1,2,i ,j) &
- v_weights(1,1,2,i ,j) * xin(2,2,i ,j) &
+ v_weights(2,2,1,i+1,j) * xin(1,1,i+1,j) &
- v_weights(2,1,1,i+1,j) * xin(2,1,i+1,j) &
+ v_weights(2,2,2,i+1,j+1) * xin(1,2,i+1,j+1) &
- v_weights(2,1,2,i+1,j+1) * xin(2,2,i+1,j+1) &
+ v_weights(3,2,1,i+1,j+1) * xin(1,1,i+1,j+1) &
- v_weights(3,1,1,i+1,j+1) * xin(2,1,i+1,j+1) &
+ v_weights(3,2,2,i ,j+1) * xin(1,2,i ,j+1) &
- v_weights(3,1,2,i ,j+1) * xin(2,2,i ,j+1)) &
* area_inv(i,j)



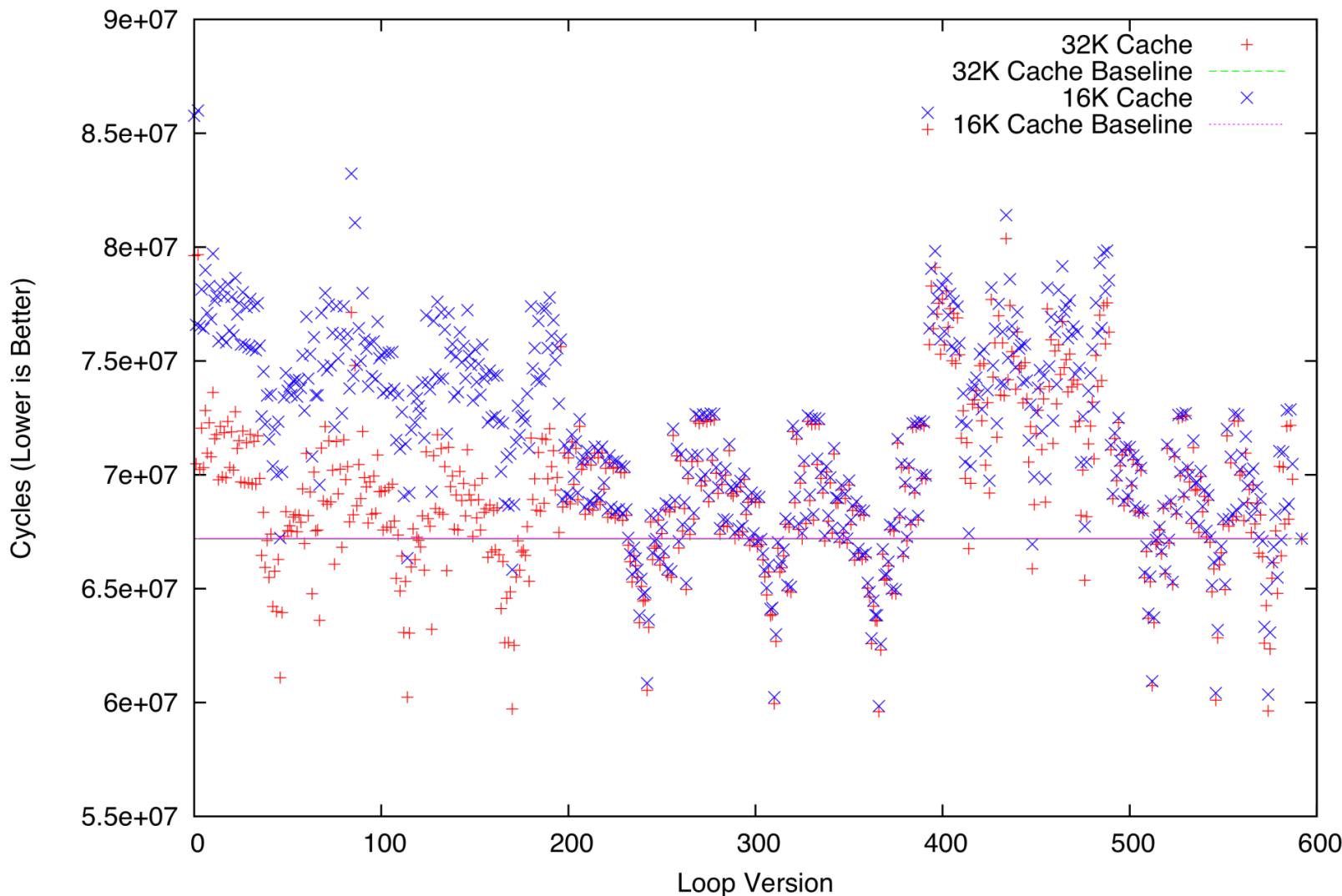
Examples

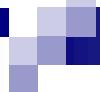
- Stencil Auto-tuner generates optimized code for these kernels on
 - Nvidia GTX280
 - Pthreads + C on Intel, AMD, and Sun Victoria Falls with 2 different indexing strategies
 - Serial Fortran



Performance Example: Serial

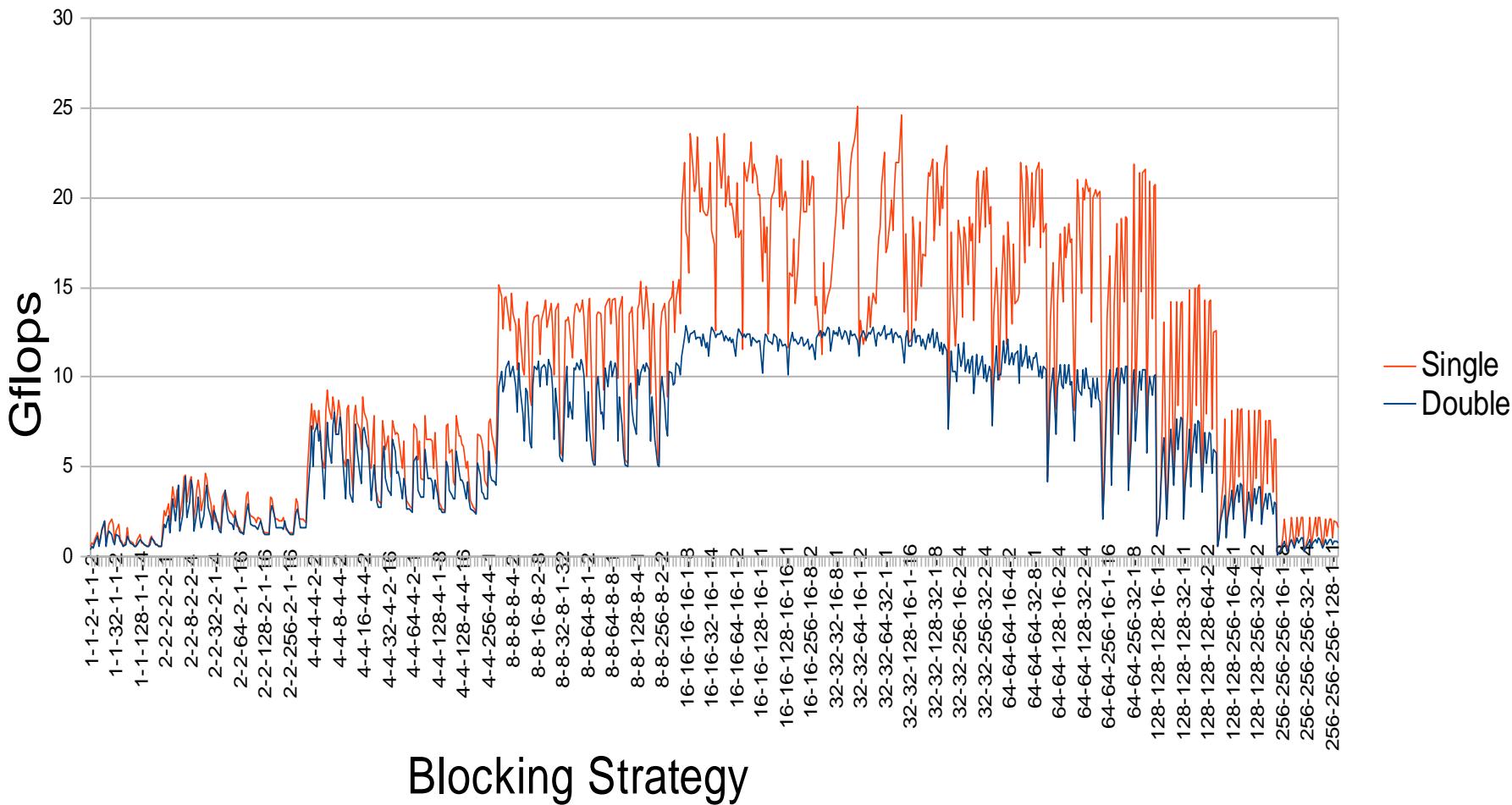
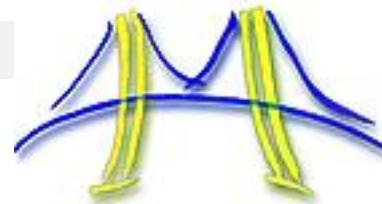
Autotuning Results for Buoyancy Loop, 16KB and 32KB Cache

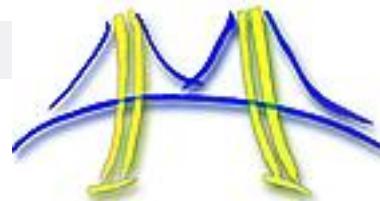
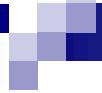




Performance Example: CUDA

CUDA Stencil Loop Performance

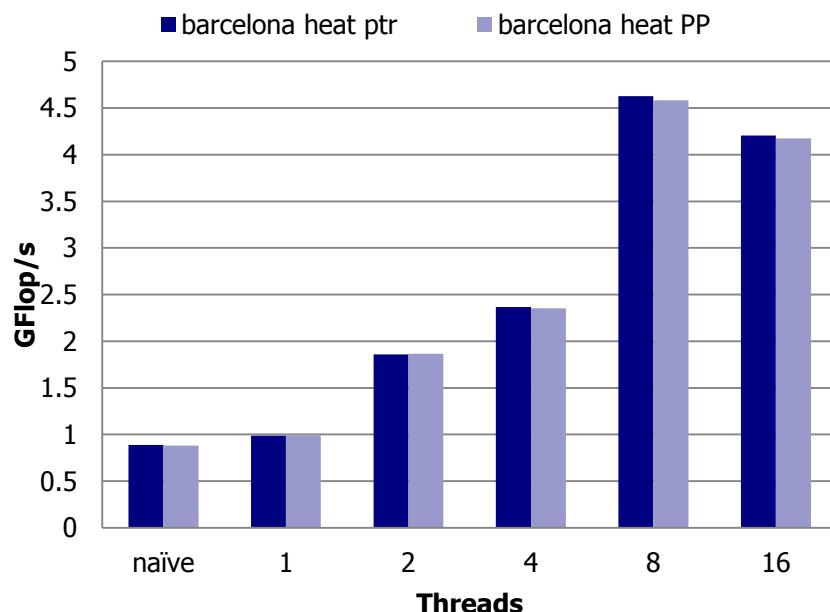




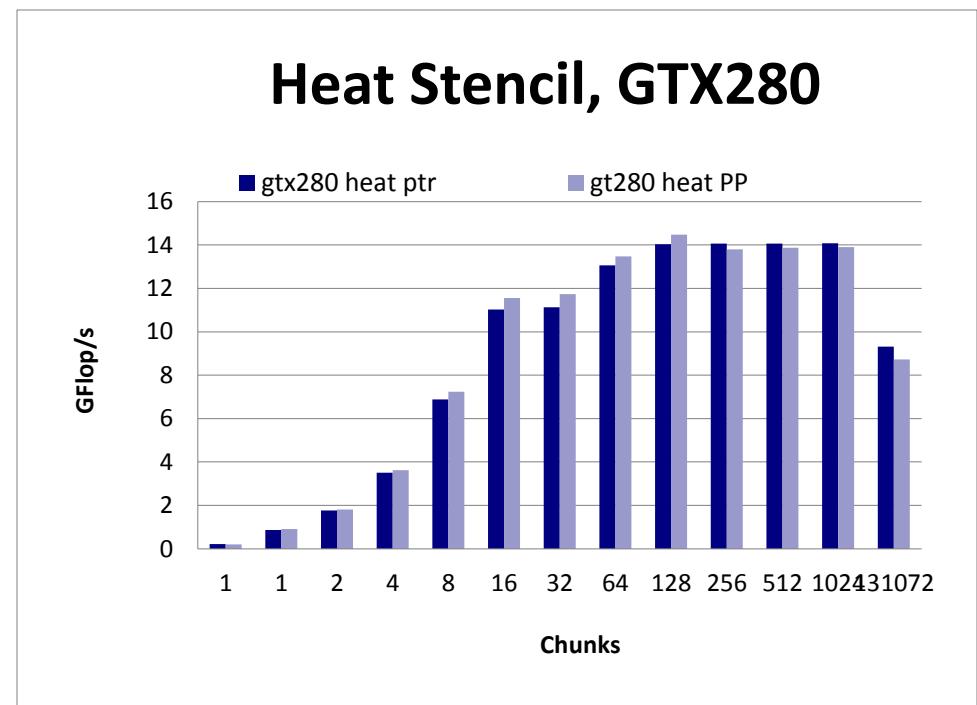
Parallel Performance Results

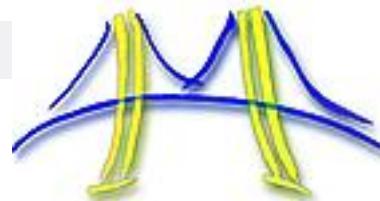
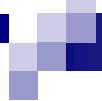
■ Heat Equation: 7pt 3D stencil from Stencil Probe

Heat Stencil, AMD 2536



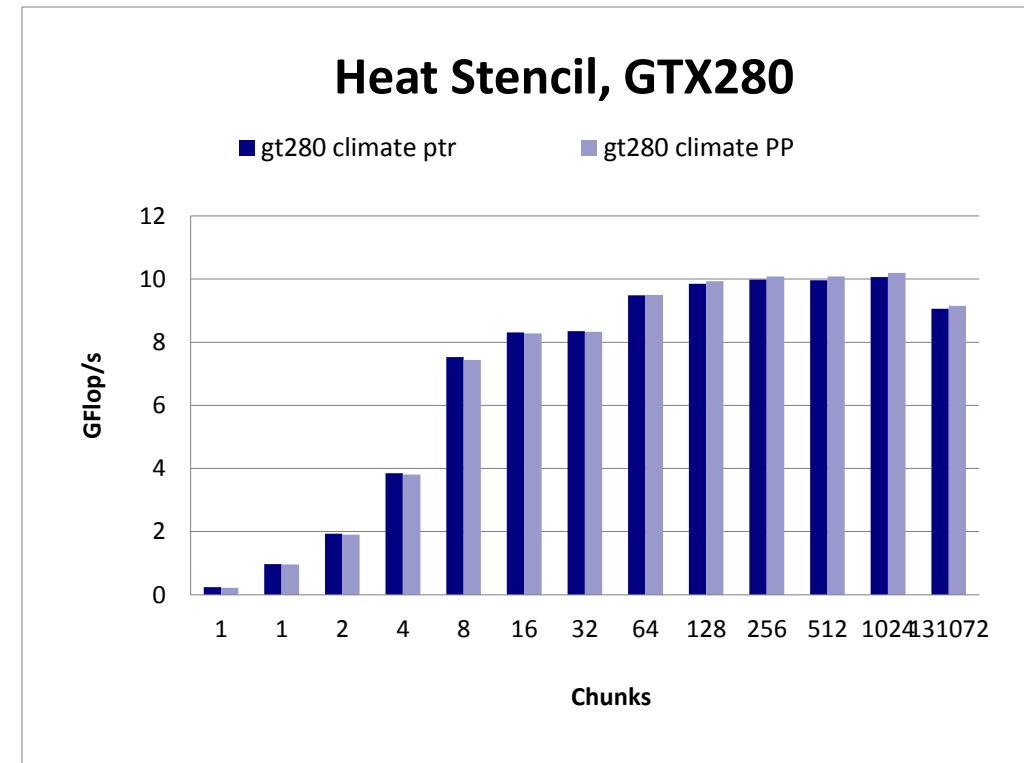
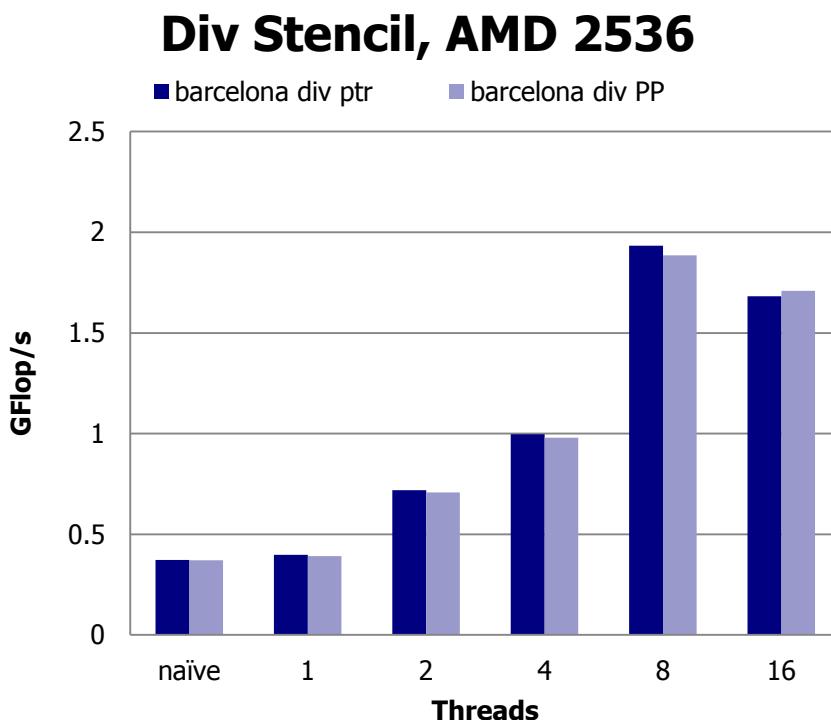
Heat Stencil, GTX280

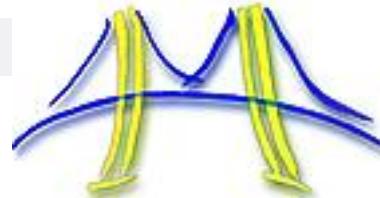
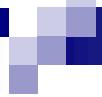




Parallel Performance Results

- 2D Divergence Kernel from Climate Application





Summary & Conclusions

- Auto-tuners should really be automatic
- Presented a framework & proof-of-concept that implements Stencil Auto-tuner
- Showed we can speed up real-world stencil kernels
- Writing Auto-tuners
 - Use better methods than string substitution
 - Make more general
 - Better search
- Performance portability for large class of stencil kernels