

Chapel With Polyhedral Transformation Using Autotuning

Tuowen Zhao and Mary Hall

The 3rd Annual Chapel Implementers and Users Workshop, 2016



Loop Transformation

- Manipulation of loop nest
 - Structure
 - Schedule
- Prior work: manually apply loop transformations in Chapel
 - I. J. Bertolacci et al. Parameterized diamond tiling for stencil computations with Chapel parallel iterators. ICS 2015
 - A. Sharma et al. Affine loop optimization based on modulo unrolling in Chapel. PGAS 2014
- We: Automatically applied loop transformations using recipes from script which enables integration with autotuning framework

Contribution

- Uses C code to capture sequential computation
- Generates Chapel programs by composing polyhedral transformations on the sequential computation and mapping from iteration spaces to Chapel domains and iterator
- Demonstrates with a simple example in Chapel the benefits of applying such transformations in conjunction with autotuning

Chapel Language

```
proc mm(A:[real],B:[real],
  an:int,ambn:int,bm:int){
  const D = {0..an-1, 0..bm-1}; // Domain
  var C : [D] real; // Domain mapped array
  forall (i,j) in D do { // Iterator
    C[i,j] = 0;
    for k in {0..ambn-1} do
      C[i, j] += A[i, k] * B[k,j];
  }
  return C;
}
```

Polyhedral Framework

- Iteration Spaces
 - A set of iteration vectors represented as integer tuples
 - Direct mapping from Chapel domain
- Transformation done by linear mapping
 - Affine loop bounds, conditional expressions, array subscripts

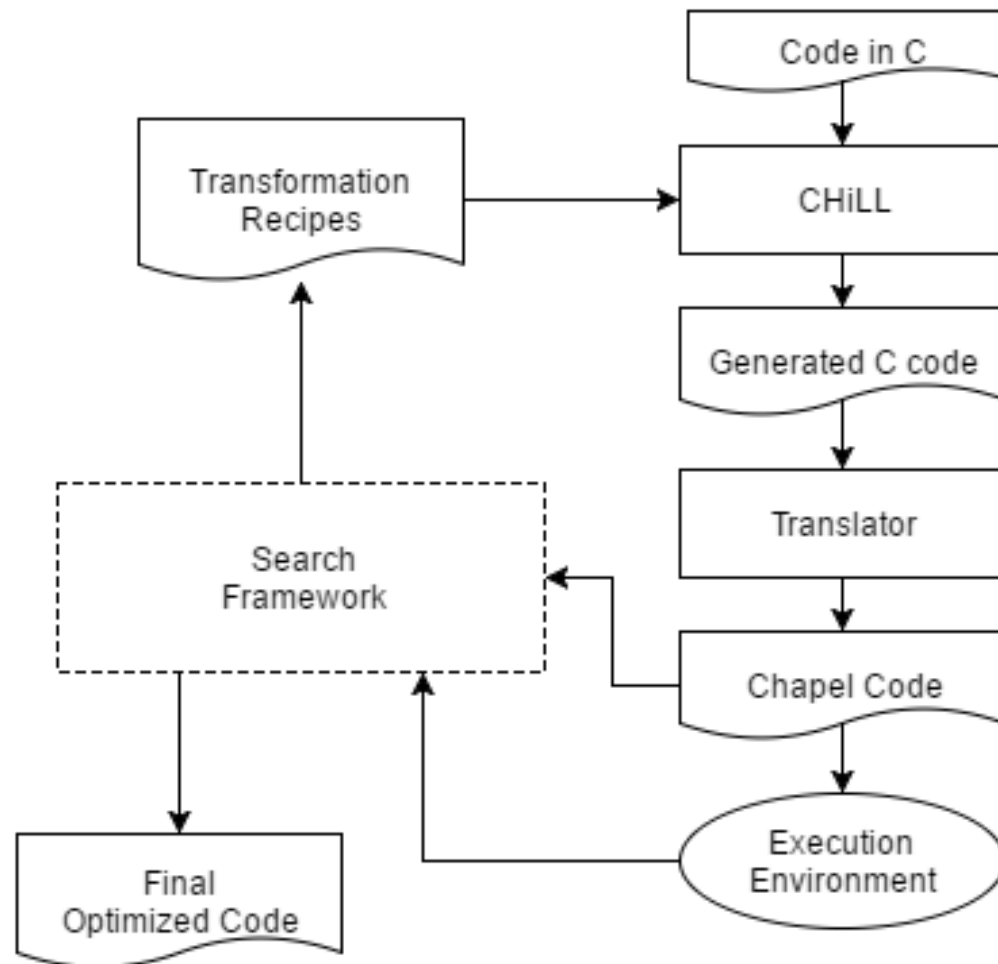
Dependence analysis

- Ensure validity of transformation and correctness of program
- Have to know the order of references to each array elements
- Cannot be applied to Chapel iterator without programmer intervention or runtime information

CHiLL

- Composable High-Level Loop transformation framework
- A polyhedral transformation and code generation framework
 - Relies on autotuning to generate highly-tuned implementations for a specific target architecture
 - Uses a transformation recipe to express optimization strategy (recipe may be generated by a compiler)

Architecture Overview



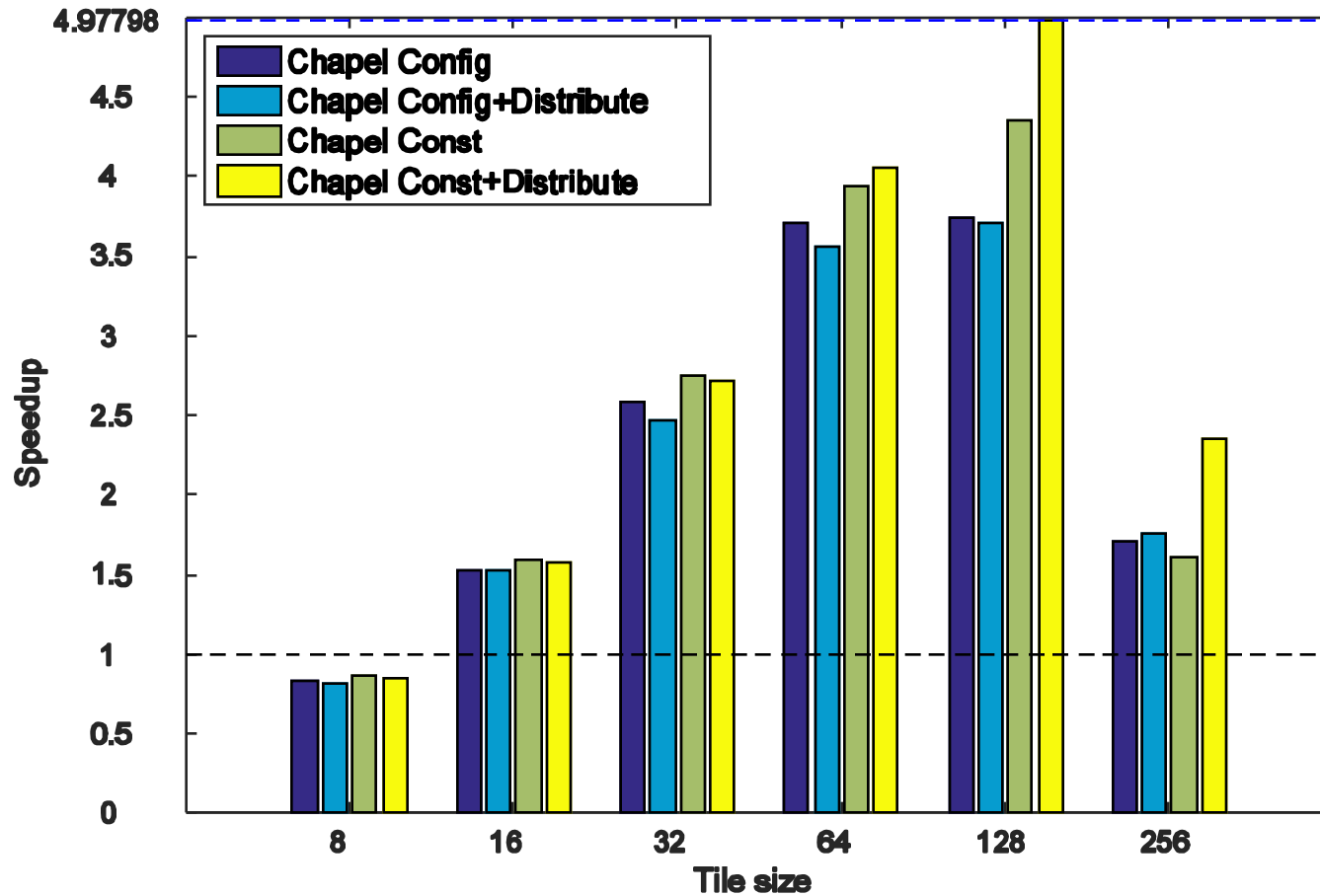
Experiment – matrix multiply

- Input in C

```
for(i = 0; i < an; i++)
for(j = 0; j < bm; j++)
{
    C[i][j]=0.0f;
    for(n = 0; n < ambn; n++)
        C[i][j] += A[i][n] * B[n][j];
}
```

- Tile sizes {8; 16; 32; 64; 128; 256}
- Distribution of the initialization code
- Tile sizes
 - Chapel's configuration variable
 - Literal constant
- Intel Haswell i7-4790K
- 16GB DDR3 RAM

Result



Stencil Computations

- Operations on structured grids
- MiniGMG
 - Geometric multigrid benchmark
 - Uses stencil computations extensively especially in smooth and residual operators
- CHiLL on MiniGMG
 - P. Basu (2015) Compiler Optimizations and Autotuning for Stencils and Geometric Multigrid. PhD thesis. University of Utah

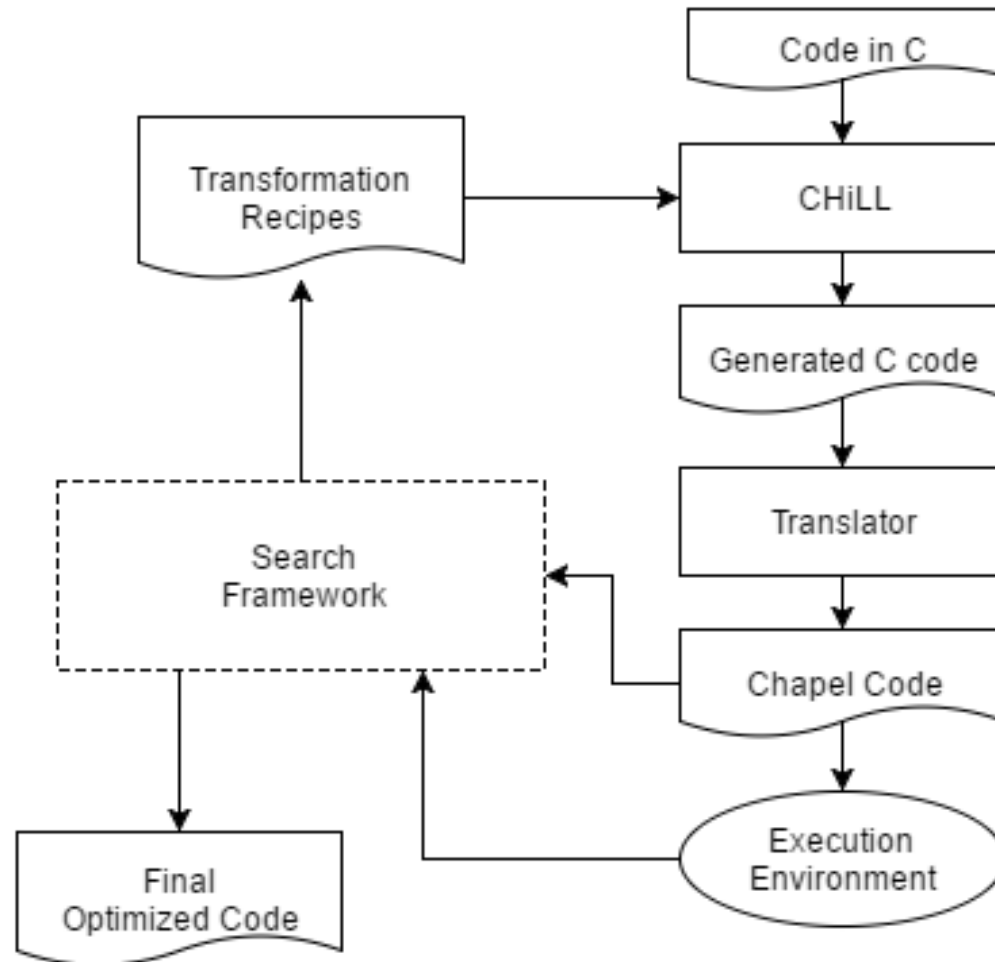
Stencil Optimizations

- Communication avoiding optimizations
 - Wavefront(loop fusing)
 - Deeper ghost zones with redundant computation

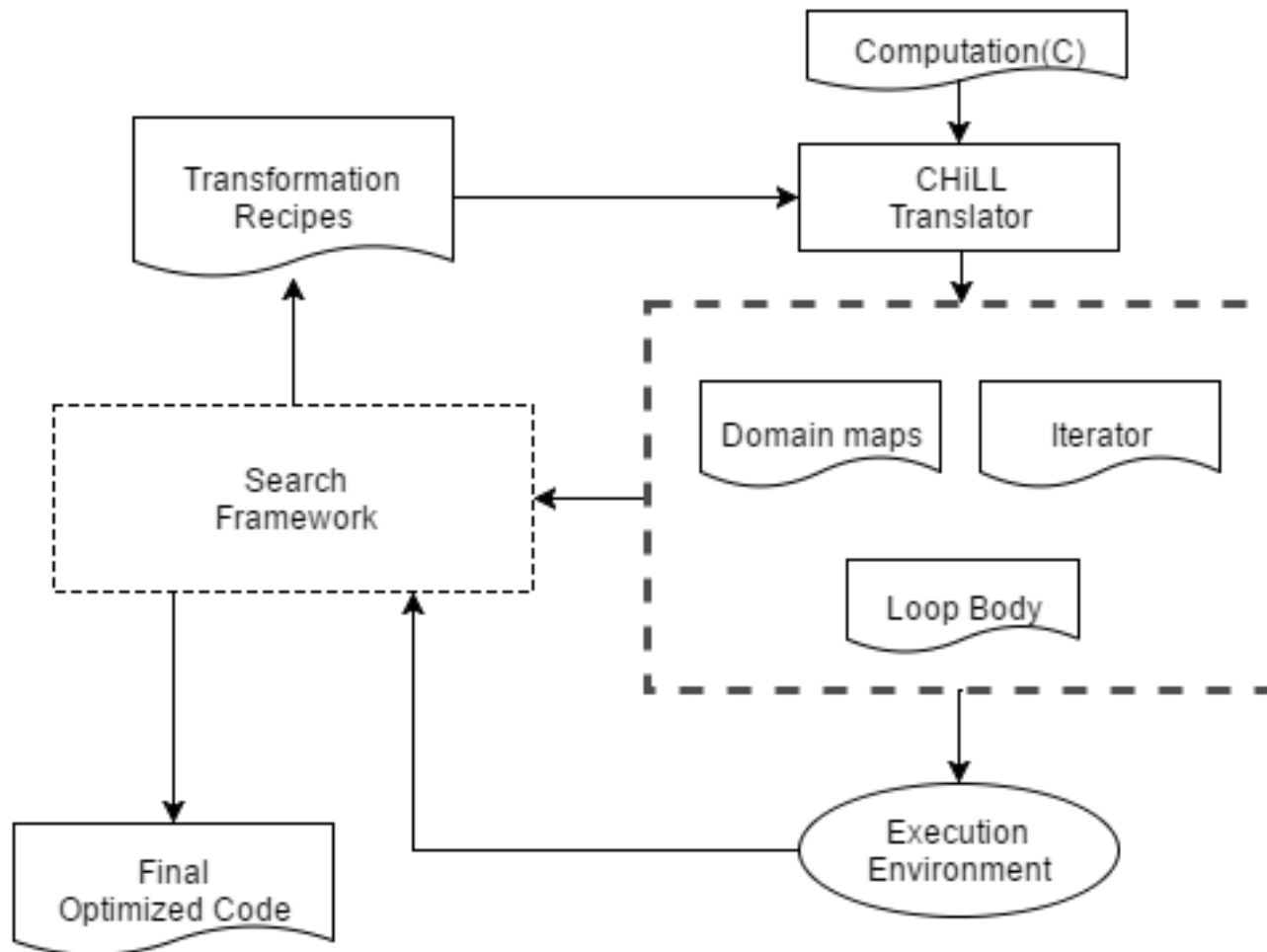
User-defined library

- StencilDist library
- Problems
 - Can't guarantee correctness(dependence)
 - Handwrite optimized code
 - Generality concern

Multi-locale Stencil



Multi-locale Stencil



Multi-locale Stencil

- Programmer writes simple serial code fragments
- Recipes provided by programmer or generated by autotuner
- Behind-the-scene generation of distributed computation and distributed data
- Produce fine-tuned code without programmer's rewriting

Conclusion

- Integrating Chapel with CHiLL
 - Instantly enables a lot of different optimization techniques that can be composed in complex sequences
 - Autotuning can be used to find the best performing combination of transformations under target architecture

Future work

- Expanding the domain of autotuning by generating and tuning domain maps and iterators
- Relaxing the transformation requirements by generalize to non-affine loop bounds and subscripts that employ indirection through an index array

Questions?