Tools for HPC



COREquations Division

Emerging Platforms

- Moore's Law
 - Power Wall
 - Memory Wall
 - ...
- Complex Parallel Systems are here
 - Core-to-Core comm./sync. only through L3 cache in AMD Barcelona and INTEL Nehalem proc. (L1/L2 private)
 - Processor-to-Processor comm./sync. even more costly in multi-socket configurations
 - Heterogenous multi-cores
 Cell BE, Host + GPUs (NVIDIA, AMD), Host + Larrabee

Programming Requirements

- Need to carefully examine
 - Data Partitioning
 - Communication
 - Synchronization

(A standard data-parallel implementation may be insufficient)

Relevant Applications/Domains

- Image/Video Processing, Vision, etc.
- Scientific/Engg./Statistical Computation
 - Computer Simulation of Physics, Signal Proc., SCADA/HMI
- Game Physics Engines
 - (SONY) Bullet Physics, (NVIDIA) Physx, (INTEL) Havoc
- CG Animation
 - Autodesk 3DS Max/Maya/XSI, Houdini, PIXAR PRMan, Mental Images. Also (open-source) Blender, Pixie & Agsis
- Bioinformatics, RNA Structure Prediction
- Geospatial Information Systems (GIS)
 - ESRI ArcGIS, PCI Geomatica. Also (open-source) GRASS 4

Application Hotspots

- Focus of Attention (Vision, Image Processing)
 - Image filtering, edge detection, feature extraction
- Motion Estimation: H.264 (Video Processing)
- Numerical Solutions to PDEs, Filters, Regression, Control & Optimization, Parameter Estimation (Sc./Engg./Stat.)
- Dynamics (Physics Simulation)
 - Fluid/Cloth/Wire/Body Simulation in Games/Movies
- Rendering (CG Animation)
 - Global Illumination/Ray Tracing
- Dynamic Programming Kernels (Bioinformatics)
- Orthorectification (GIS)

Hotspot Characteristics

- Highly Compute-Intensive
- High Data Volume
- Very well structured kernel fragments

Possible Solutions

- Automatically generate libraries
- Develop highly tuned libraries

Automatic Library Generation (The COREquations Engine)

Automatic Library Generation

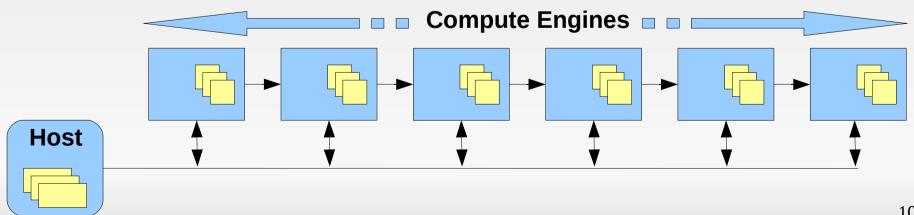
- Domain Specific Compiler Engine
- Class of Problems: The Polyhedral Model
 - Incorporates most parts of four dwarfs in the "Berkeley View*" (each was expected to require an independent technology)
 - Structured Grids
 - Dense Matrix
 - Most of Dynamic Programming
 - Some of Graphical Models
- * The Berkeley View is a recent study by application/compiler/hardware specialists that outlines the most important problem domains for parallel platforms

The Polyhedral Model

- Most dynamic instr. from iterative kernels
- Manual code optimization is error prone and time consuming
- We wish to develop equivalent high level compiler optimizations
 - Analyze program behavior
 - Design custom transformations
 - Guarantee validity

Parallel Code Generation

- Host owns data (partitions, communicates & synchronizes)
- "Compute Engines" process data
- Host reports results back to user
- Heterogeneous threads



Specifications

- Single Assignment Loops (actually equations)
- Matrix Multiplication (The HPC "Hello World")

```
affine matrix product {P, Q, R | P>0 && Q>0 && R>0} Program Parameters
  given float A \{i,k \mid 0 \le i \le P \&\& 0 \le k \le Q \};
                                                                   Input Variables
         float B \{k,j \mid 0 \le k \le 0 \le j \le R \};
returns float C {i,j,k | 0<=i<P && 0<=j<R && k==Q-1 }; Output Variable
        float temp C \{i,j,k \mid 0 \le i \le P \&\& 0 \le j \le R \&\& -1 \le k \le Q \};
using
                                                                    Local Variable
through
temp C[i,j,k] = case
                                                          Equation for Accumulator
                       { | k \ge 0} : temp C[i,j,k-1]+(A[i,k]*B[k,j]);
                           k==-1 : 0;
                   esac;
C[i,j,k] = temp C[i,j,k];
                                                               Equation for Result
```

Script-Driven Compilation

```
source("Setup.bsh");
```

Read the program

ConnectServer("localhost");

ReadProgram("matrix_product.alphabets");

SetMemoryMap("matrix_product","C",

Desired memory layout of output "(P,Q,R,i,j,k->P,Q,R,i,j)","0,0");

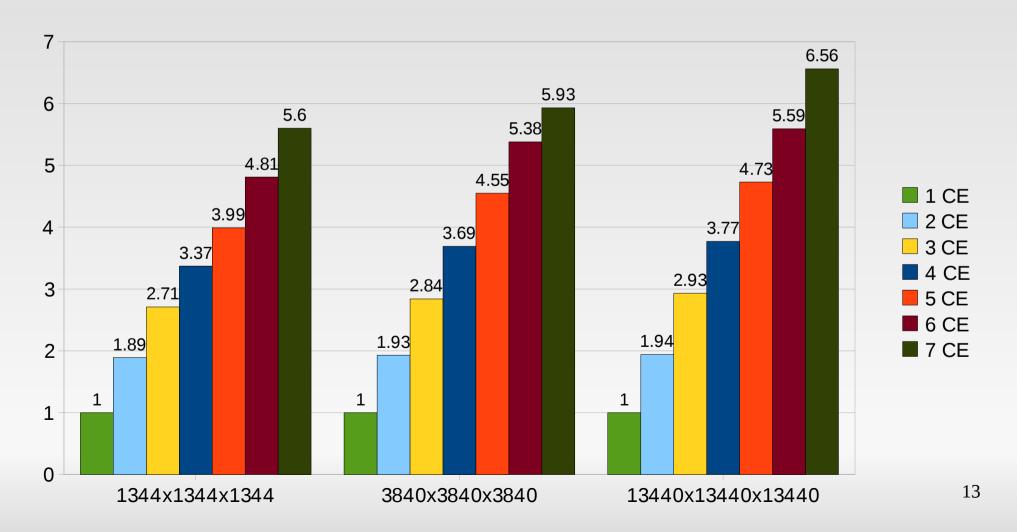
MPICodeGen(8,"32,32,32");

Generate code for 8 processors, tiling for parallelism with sizes 32x32x32 along i,j and k

Memory Maps are only needed for output variables Tool finds <u>optimal memory layout</u> for local variables

Performance I: Scaling

Near-Perfect Scaling



Performance II: Raw Comparison

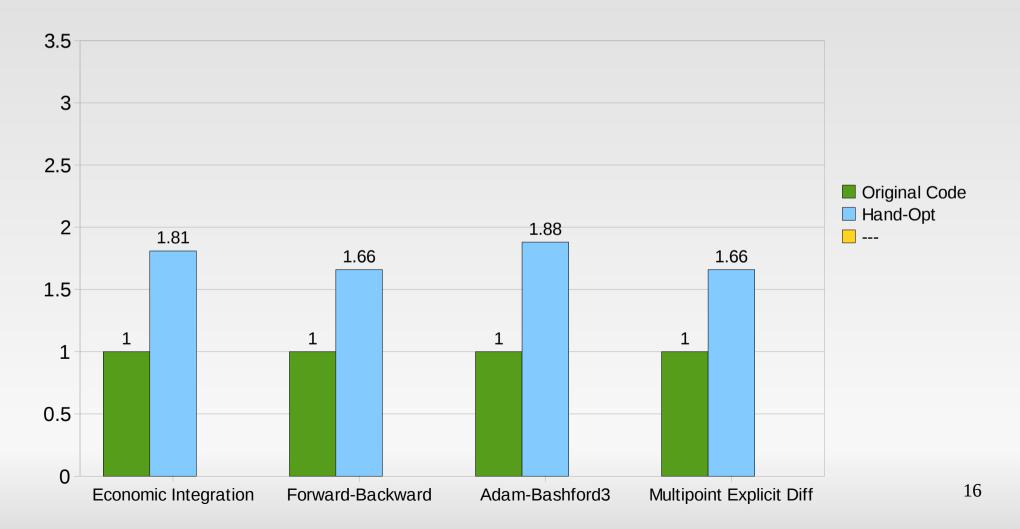
- Compare to Hand-Optimized Sequential Code
- Before, our MPI Code Generator, we wanted to generate Good Sequential Code

Sequential Code Generation

- Case Study: PDE Solver from Atmospheric Sc.
 - 2D-3D stencil. Five-point update
 - Over time, elements on an area update their value based on previous value and North, South, East and West neighbors
 - Periodic boundary conditions
 - West border depends on East border, etc.
 - SW corner depends on SE and NW corners
- Hand-Optimized by Nathan Burnett, a Masters Student at CSU
 - Performed optimizations such as loop interchange, fusion, fission, unrolling, removal of lookup-tables ...

Performance (Hand-Optimization)

Intel CORE2 Duo (2.3Ghz)



EI PDE Solver

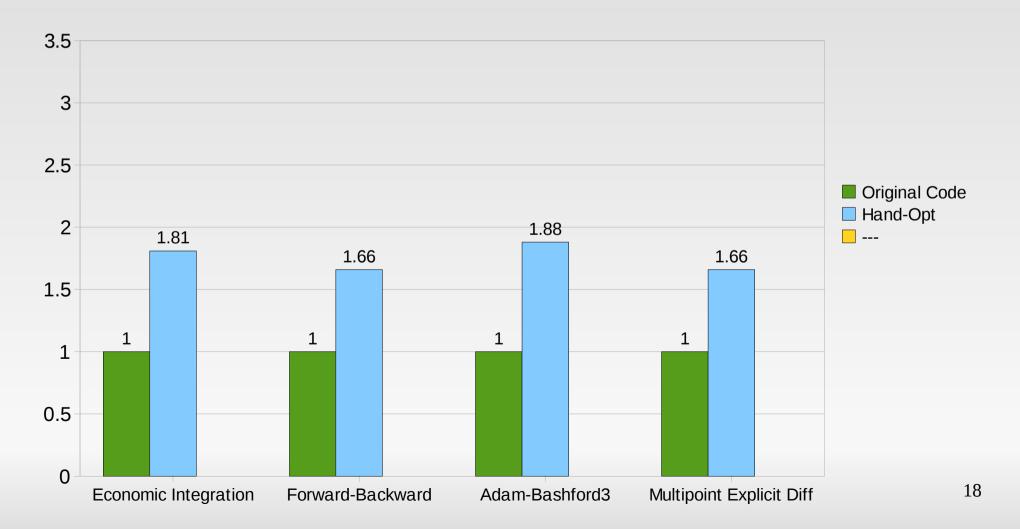
results[j,i] = h[i,j,TMAX];

```
through
h[i,j,k] = case
                                                                             Main Variable
                {| k==0}: h init[j,i];
                \{ \mid 0 < k \} : h[i,j,k-1] - delt * H0 * delta[i,j,k-1];
            esac;
lap h[i,j,k] =
                                                                           Key Computation
 case
  \{ | i==0 \&\& j==0 \} : (h[i+1,j,k] + h[2im h-1,j,k] + h[i,j+1,k] + h[i,2jm h-1,k] \}
  SW Corner
                               -4*h[i,j,k]) * inv d sq;
  \{ | i==0 \&\& 0 <= j < 2jm h \} : (h[i+1,j,k] + | h[2im h-1,j,k] + h[i,j+1,k] + h[i,j-1,k] \}
  West Border
                                   -4*h[i,j,k]) * inv d sq;
  \{ \mid 0 < i < 2im \ h-1 \& \& 0 < j < 2jm \ h-1 \}: (h[i+1,j,k] + h[i-1,j,k] + h[i,j+1,k] + h[i,j-1,k]
  Main Body
                                    -4*h[i,j,k]) * inv d sq;
 esac;
delta[i,j,k] = case
                                                                           Auxillary Variable
                    \{ | k==0 \} : 0;
                    \{ \mid 0 < k \} : delta[i,j,k-1] - delt * g * lap h[i,j,k];
                 esac;
```

Output

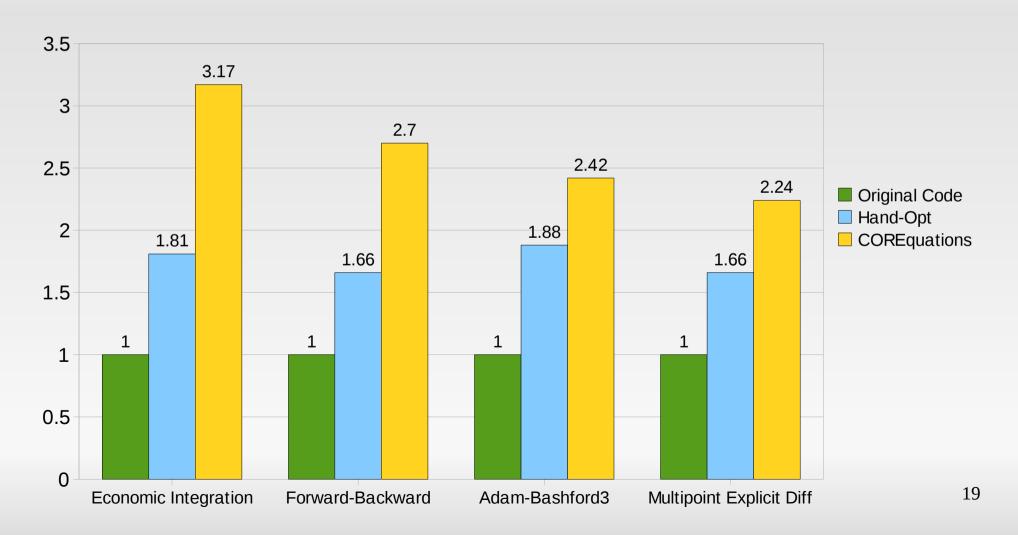
Performance (Hand-Optimization)

Intel CORE2 Duo (2.3Ghz)



Then, I provided code for his equations

Intel CORE2 Duo (2.3Ghz)



Why is generated code so good?

- Memory Optimization
 - (Prefetch-friendly) Alignment
 - Storage Minimization
- Loop Optimizations (just much more of them): Automatic loop-generator specializes code for all boundary cases
 - Aggressive Fusion
 - Aggressive Fission

Triangular Matrix Multiplication

- Beyond "Hello World"
- P=Q=R (=N say). Add triangular constraints

```
affine triangular matrix product {N N>0}
  given float A \{i,k \mid 0 \le i \le N \&\& 0 \le k \le i \};
          float B \{k, j \mid 0 <= k < N \&\& k <= j < N \};
returns float C {i,j,k | 0 <= i <= j \&\& 0 <= j < N \&\& k == i } | |
                    \{i,j,k \mid 0 \le i \le N \&\& 0 \le j \le i \&\& k == j \};
using float temp C {i,j,k | 0 \le i \le N \&\& 0 \le j \le N \&\& -1 \le k \le (i,j) };
through
temp C[i,j,k] = case
                        { | k \ge 0} : temp C[i,j,k-1]+(A[i,k]*B[k,j]);
                        \{ \mid k==-1 \} : 0;
                    esac;
C[i,j,k] = temp_C[i,j,k];
```

Triangular Matrix Multiplication

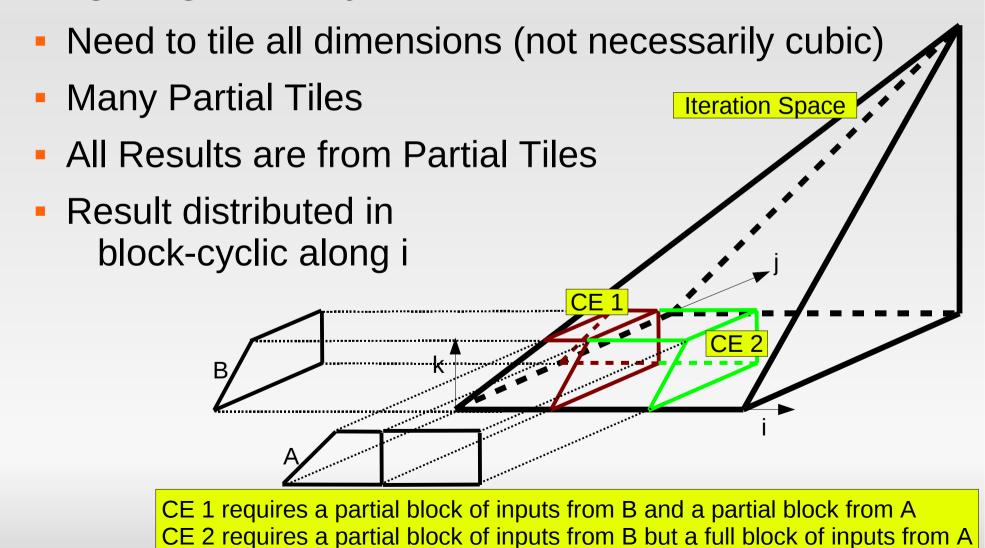
Sequential Code

```
for i = 0 to n-1 {
         for j = 0 to n-1 {
           C[i,j] = 0;
                                                Iteration Space
            for k = 0 to min(i,j) {
              C[i,j] += A[i,k]*B[k,j]
Why write equations when you can write code?
   Iteration Space
     is a polyhedron
```

 Now, let us generate MPI code (What is each Compute Engine iterating over?)

Triangular Matrix Multiplication

Targeting memory-constrained architectures

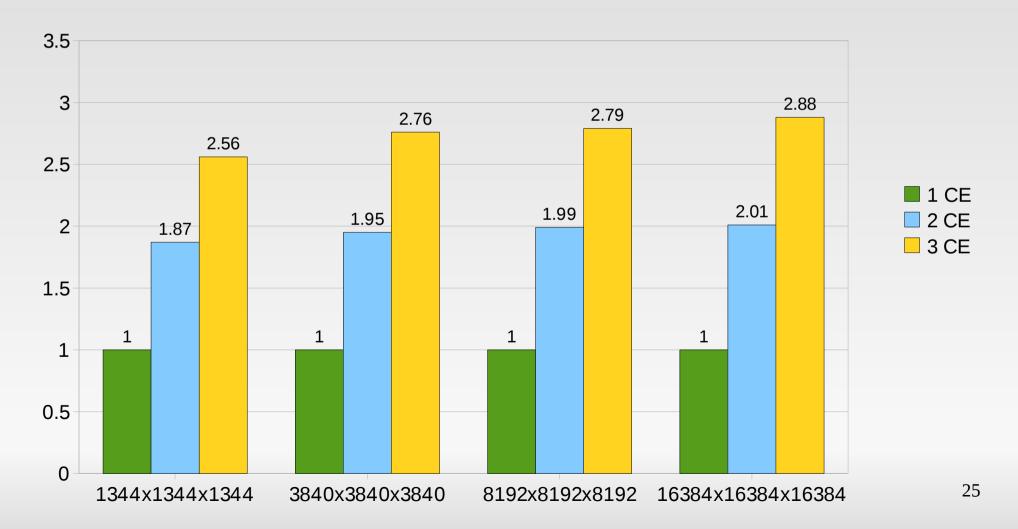


In Short ...

- Writing MPI code even for a straightforward specialization of Matrix Multiplication is extremely time consuming and error-prone
- In COREquations, just run the same script for the new specification

Performance

Again, Near-Perfect Scaling



Design Space Exploration

Sequential Code for Matrix Multiplication

```
for i = 0 to n-1 {
    for j = 0 to n-1 {
        C[i,j] = 0;
        for k = 0 to N-1 {
            C[i,j] += A[i,k] * B[k,j]
        }
        Subsequent reads of A are along the inner index.
        Memory accesses to A are prefetch friendly.
```

Subsequent reads of B are along the outer index. Memory accesses to B are not prefetch friendly.

What about writes to C?

Writes to C are to the same location throughout the inner loop

Design Space Exploration

- A number of choices in code-generation
 - Tiling: (shape and size)
 FixedTiledCodeGen("48,32,63");
 - Different Memory Layouts of Variables
 SetMemoryMap("matrix_product","B",
 "(P,Q,R,k,i->P,Q,R,i,k)","0,0");
 - Permutation of Loops
 - With j as the inner loop, A and C are accessed only once along the inner loop. B[k,j] becomes Prefetch-friendly.

```
CoB("matrix_product","C","(P,Q,R,i,j,k->P,Q,R,i,k,j)");
CoB("matrix_product","temp_C","(P,Q,R,i,j,k->P,Q,R,i,k,j)");
^{27}
```

Future Work

- We're going to make it easier to program
 - Higher Level Language
 - Matrix Product should simply be specified as C[i,j] = SUM([k], A[i,k]*B[k,j]);
 - Domain Inferencing
- More Analysis (Scheduling, etc.)
- Get the Human-Out-of-the-Loop

 Porting to Newer Architectures (GPUs, Cell, Larrabee, ...)

Summary

- For specifications in Alphabets
 - MPI is easy
 - Design Space Exploration is easy
 - Scalable Performance
- Auto-verification
 - Compiler-assisted programming to avoid double-or incomplete-definitions, out-of-bound references ...
 - Major source of errors is eliminated

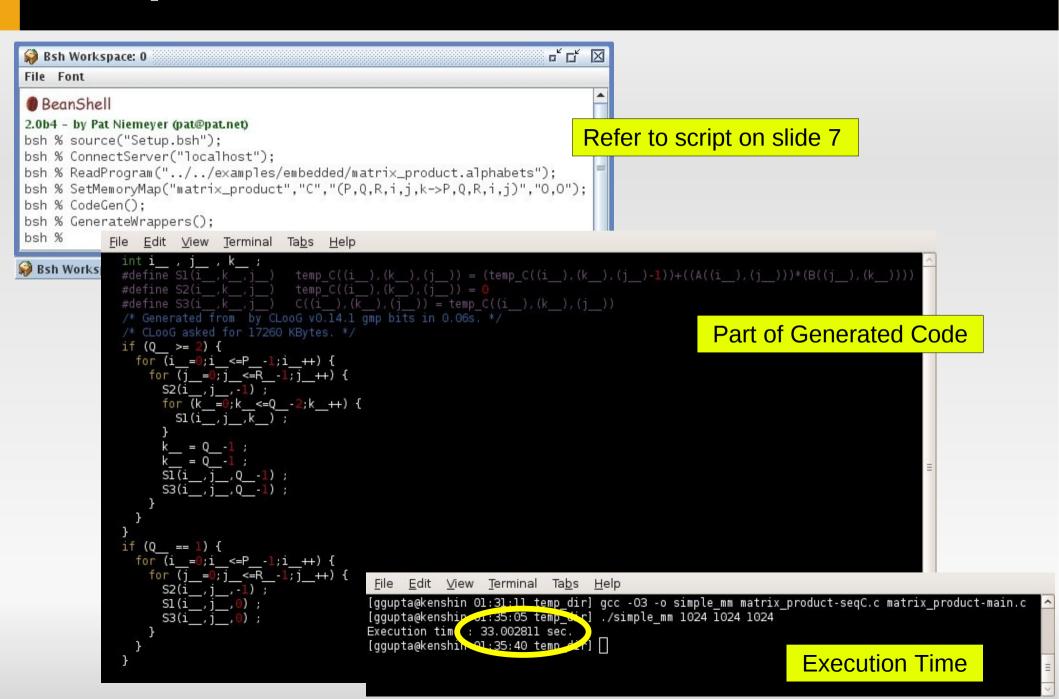
Generality of Solution

 A new language for generating HPC code for compute-intensive kernels, but also ...
 An Intermediate Representation for optimization of loops in conventional languages

Appendix

Inexpensive Design Space Exploration

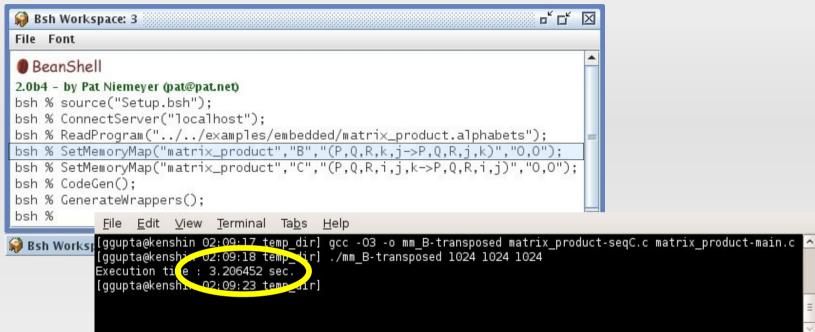
Simple Matrix Product



Tiled Matrix Product

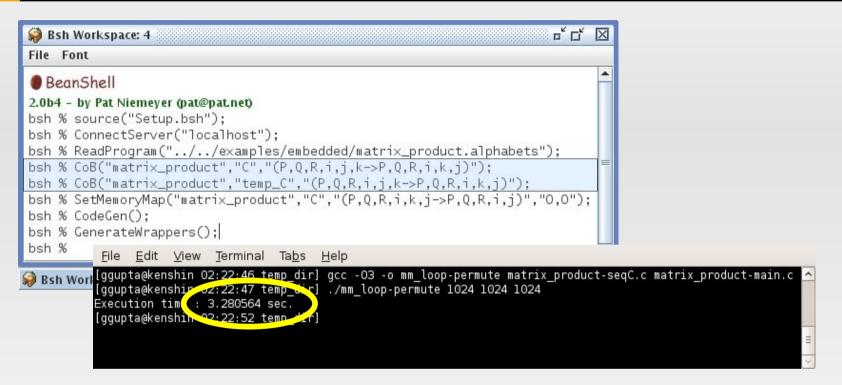
```
- T 🗵
Bsh Workspace: 2
File Font
 BeanShell
2.0b4 - by Pat Niemeyer (pat@pat.net)
bsh % source("Setup.bsh"):
bsh % ConnectServer("localhost"):
bsh % ReadProgram("../../examples/embedded/matrix_product.alphabets");
bsh % AShow():
                                                                 Loaded Program
affine matrix_product {P,Q,R | P>=1 && Q>=1 && R>=1}
aiven
   float A {i,k | k>=0 && i>=0 && 0-k>=1 && P-i>=1}:
   float B {k, j | j>=0 && k>=0 && R-j>=1 && Q-k>=1};
returns
   float C {i,j,k | Q-k==1 && j>=0 && i>=0 && R-j>=1 && P-i>=1};
usina
   float temp_C {i,i,k | k>=-1 && i>=0 && i>=0 && R-i>=1 && 0-k>=1 &&
P-i>=1}:
through
temp_C[i,j,k] = case
   \{ | k \ge 0 \} : (temp_C[i,j,k-1] + (A[i,k]*B[k,j]));
    \{ \mid k==-1 \} : 0;
esac;
C[i,j,k] = temp_C[i,j,k];
bsh % SetMemoryMap("matrix_product", "C", "(P, Q, R, i, j, k->P, Q, R, i, j)", "0, 0");
bsh % FixedTiledCodeGen("32,32,32");
bsh % GenerateWrappers();
                                          File Edit View Terminal Tabs Help
bsh %
                                         [qqupta@kenshin 01:53:05 temp dir] qcc -03 -o tiled mm matrix product-fixedtiled.c matrix product-main.c -lm
                                         [ggupta@kenshir 01:53:08 temp_dir] ./tiled_mm 1024 1024 1024
Bsh Workspace: 2
                                         Execution time: 15.598294 sec.
                                         [ggupta@kenshim 91:53:25 temp_ar]
```

Changing the memory layout of B



- Significant performance upgrade by having a prefetch-friendly layout of the input array B
- Can we obtain similar performance with the standard memory layout for B

Permuting loop indices



 By having j as the innermost loop index, accesses (to arrays A and B) preserve locality