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# Performance, Portability, and Dreams

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# Why Performance Portability?

- HPC is Performance
- A big part of the programming crisis is caused by the challenge of obtaining performance (even) on a single platform
  - This is an unsolved problem
    - Easy example: Implement a barrier. Communicates a single bit of information.
      - Easy to write a simple implementation (e.g., use an atomic counter).
      - Efficient implementations require clever algorithms, attention to memory hierarchies, special instructions, and are still publishable
        - “Algorithms for Scalable Synchronization on Shared-Memory Multiprocessors”, ACM TOCS, 1991
        - Recent results such as “Efficient Barrier Implementation on the POWER8 Processor”, HiPC 2015
  - Performance portability related to programming productivity
    - And a harder problem that is getting even harder

Platform-specific

# Why Is Performance Portability so Hard

- Its been hard
  - Predicting performance for a single system is very difficult
    - Systems are complex
    - Behavior has random elements
    - Interactions between parts is hard to predict
- After more than 20 years of stability, processor architectures are diversifying and changing
  - More types of systems – e.g., vectors/streams in GPUs
  - Rapid innovation – new instructions, memory architectures, ...
  - Effective\* use of these requires someone to adapt to the differences
    - Please make it someone else!
- Even if it is someone else
  - Many costs and risks to maintaining multiple versions

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# Tradeoffs

- Implicitly, performance portability is intended to reduce the amount of work needed to achieve adequate performance to meet the needs that the computing serves
- How much (programmer re-) work is acceptable to achieve performance portability?
- What constraints or other limitations are acceptable?
  - Choices of data structure, code complexity, reproducibility, compile time, sensitivity to changes in input data, ...

# What is the Problem Statement?

- General case (strong performance portability) – get the fastest solution to the problem on all systems – is far too hard – requires picking model, algorithm, data structure, and implementation
  - Best algorithm/data structure choice often unknown
  - Algorithm may depend on platform
    - Proof – parallel algorithms that trade less synchronization against more work vs. sequential algorithms
- Given a family of data structures and an algorithm, choose the data structure instance and implementation
  - E.g., Array index ordering; Structure of arrays vs array of structures vs structure of arrays of structures; sparse matrix ordering
- Given a data structure and an algorithm, generate “good” code that performs well
  - Problem: choices here are important for performance
  - Problem: Still hard even in simple cases
  - Problem: “Extra” semantics in language – e.g., order of operations for floating point, e.g. in dense matrix-matrix multiply, can limit options even if not intended by programmer

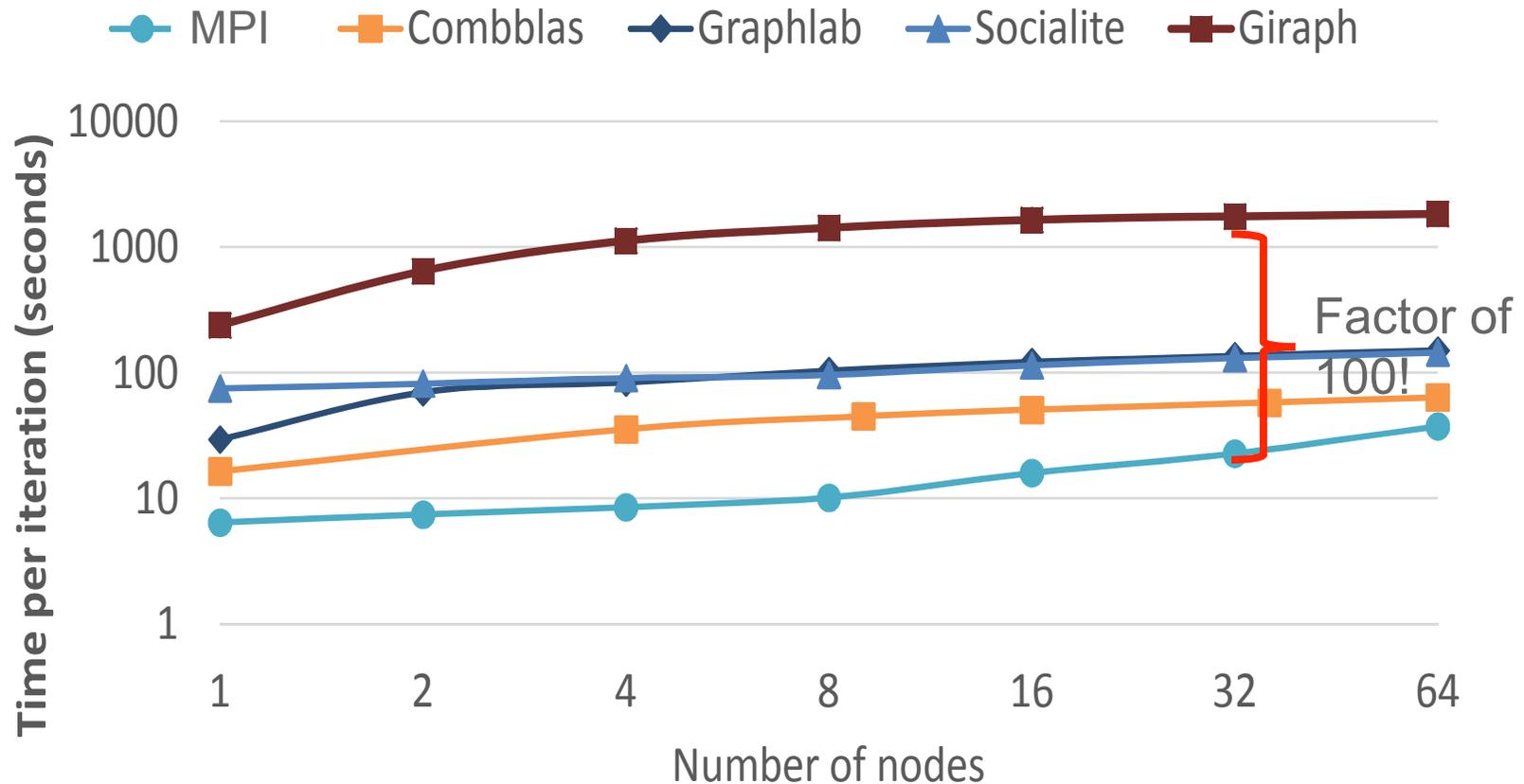
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# Metrics for Success

- What is success?
- Need to quantify both portability and performance
- Should include impact on productivity
  - A performance portable code that is no longer maintainable or that is too brittle for further development is probably not an improvement
- Not an easy linear function
  - Different users and communities may choose different weights for the metric

# Productivity and Performance

Collaborative Filtering (Weak scaling, 250 M edges/node)



**Navigating the Maze of Graph Analytics Frameworks using Massive Graph Datasets**

Nadathur Satish, Narayanan Sundaram, Md. Mostofa Ali Patwary, Jiwon Seo, Jongsoo Park, M. Amber Hassaan, Shubho Sengupta, Zhaoming Yin, and Pradeep Dubey

# Dangers in Performance Portability

- One easy way – make all performance mediocre
  - One vendor did this in the '80s with their vector hardware, to avoid too large a variability in performance
    - Goal was no performance “surprises”
  - Related – predictable performance – a goal and elegant feature of BSP (Bulk Synchronous Programming)
    - How much opportunity for higher but less predictable performance are you willing to give up for predictable performance? Do your users agree with you?
- Another easy way – claim that it can be reduced to a previously solved problem
  - E.g., Claim the compiler can take care of it
  - This is a fantasy
- We clearly need a good definition...

# One Definition

- An application is performance portable if it achieves a **consistent ratio** of the actual time to solution to either the best-known or the theoretical best time to solution on each platform with **minimal platform specific code** required.
  - From <http://performanceportability.org/perfport/definition/>
  - Note that other definitions are mentioned with different focus and levels of precision
- “Best-known” time to solution is a big loophole
  - For a new system, best known is your own best time
  - If there is only one code, and it runs and there is no theoretical best time, the code is performance portable, regardless of the actual performance
    - That consistent ratio is 1 😊
- See more on one view of performance portability at
  - <http://performanceportability.org/>

# What Is Performance Portability?

- Is it:
  - A code is performance portable if it achieves at least 100-X% of the achievable performance on all platforms
- Do I need to add constraints?
  - with the same algorithm
  - and the same data structures
  - and the same input and output data organization and format
  - and the same build system (e.g., makefile)
- How large can X be for this definition to be useful?
  - 1? 10? 50? 99? 99.99999?
- Is X the same for all platforms?
  - Alternately, is there an *absolute* performance target, and the code is performance portable if the code meets or exceeds that performance on all platforms of interest?
- Is there a scaling of X based on the cost (\$) of the platform?

# Defining Performance Portability

- And what about the correctness constraints
  - Is the output strongly or weakly deterministic?
  - Is bitwise identical output required?
- What is the definition of achievable performance?
  - FLOPS?
  - FLOPS and Memory Bandwidth (“roofline”  
<https://dl.acm.org/citation.cfm?id=1498785> )
  - FLOPS and Memory Bandwidth and Latency (Execution-Cache-Memory (ECM) model  
[https://link.springer.com/chapter/10.1007%2F978-3-642-14390-8\\_64](https://link.springer.com/chapter/10.1007%2F978-3-642-14390-8_64) )
  - FLOPS and Memory Bandwidth and Instruction Rate (“Achieving high sustained performance in an unstructured mesh CFD application”  
<https://dl.acm.org/citation.cfm?id=331600> , 1999)

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# What Is Performance Portability?

- Is it:
  - A code is performance portable if it runs with acceptable performance without any source code change (or architecture-specific directives) on the platforms of interest
- This is squishy. What is
  - Acceptable performance
  - Without any source code change
  - Platforms of interest
- What if I make this *more* squishy
  - A code is performance portable if it runs with acceptable performance with no onerous source code or build system changes on most of the platforms of interest

# Some Performance Portability Questions

- “How much performance would you be willing to give up by replacing the two optimal applications by a single one?”
  - <https://software.intel.com/en-us/blogs/2017/03/30/rainbows-unicorns-and-performance-portability> (Robert Geva, Intel)
- How much are you willing to spend to achieve performance portability
  - E.g., if maintaining two codes takes 100 FTE/each and recasting a code in a new system takes 250 FTE, is that acceptable? What if it costs 2500 FTE?
- These ask *quantitative* questions about performance portability
- They also get to the heart of *why* someone might want performance portability

# Some Different Approaches to Performance Portability

- Language based
  - Existing languages, possibly with additional information
    - Info from pragmas (e.g., align) or compile flags (assume associative)
  - Extensions, especially for parallelism
    - Directives + runtimes, e.g., OpenMP/OpenCL/OpenACC
    - May also relax constraints, e.g., for operation order, bitwise reproducibility
  - New languages, especially targeted at
    - Specific data structures and operations
    - Specific problem domains
- Library based (define mathematical operators and implement those efficiently)
  - Specific data structure/operations (e.g., DGEMM)
  - Specific operations with families of data structures (e.g., PETSc)
    - This is likely the most practical way to include data-structure and even algorithm choice
    - At the cost of pushing the performance portability problem onto the library developers

# Some Different Approaches to Performance Portability

- Tools based
  - Recognize that the user can always write poorly-performing code
  - Support programming in finding and fixing performance problems
  - Example: Early vectorizing compilers gave feedback about missed vectorization opportunities; trained programmer to write “better” code
- Programmer support and solution components
  - Work with programmer to develop code
  - Source-to-source tools to transform and to generate code under programmer guidance
  - Autotuning to select from families of code
  - Database systems to manage architecture and/or system-specific derivatives
- Magic
  - Any sufficiently advanced technology is indistinguishable from magic. (Clarke’s 3rd law)
  - Any sufficiently advanced technology is indistinguishable from a rigged demo.
- Note these approaches are not orthogonal
  - Successful performance portability requires many approaches, working together

# “Domain-specific” languages

- (First – think abstract data-structure specific, not science domain)
- A possible solution, particularly when mixed with adaptable runtimes
- Exploit composition of software (e.g., work with existing compilers, don't try to duplicate/replace them)
- Example: mesh handling
  - Standard rules can define mesh
    - Including “new” meshes, such as C-grids
  - Alternate mappings easily applied (e.g., Morton orderings)
  - Careful source-to-source methods can preserve human-readable code
  - In the longer term, debuggers could learn to handle programs built with language composition (they already handle 2 languages – assembly and C/ Fortran/...)
- Provides a single “user abstraction” whose implementation may use the composition of hierarchical models
  - Also provides a good way to integrate performance engineering into the application

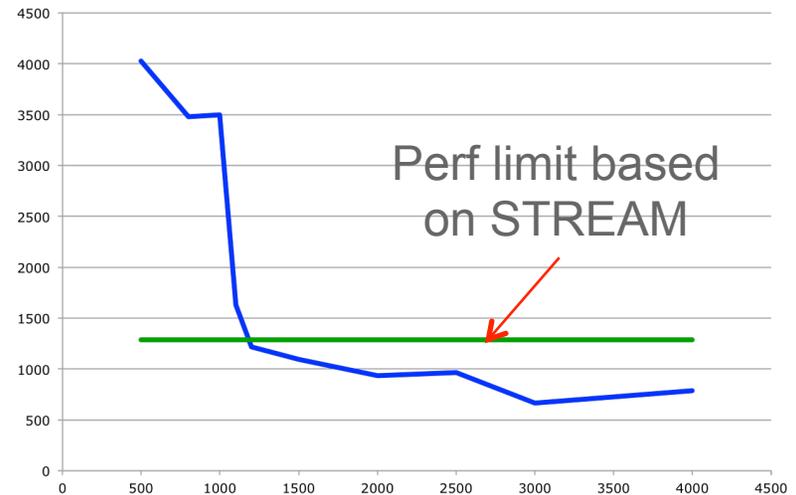
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# Let The Compiler Do It

- This is the right answer ...
  - If only the compiler *could* do it
- Lets look at one of the simplest operations for a single core, dense matrix transpose
  - Transpose involves only data motion; no floating point order to respect
  - Only a double loop (fewer options to consider)

# A Simple Example: Dense Matrix Transpose

- do j=1,n  
  do i=1,n  
     $b(i,j) = a(j,i)$   
  enddo  
enddo
- No temporal locality (data used once)
- Spatial locality only if (words/cacheline) \* n fits in cache

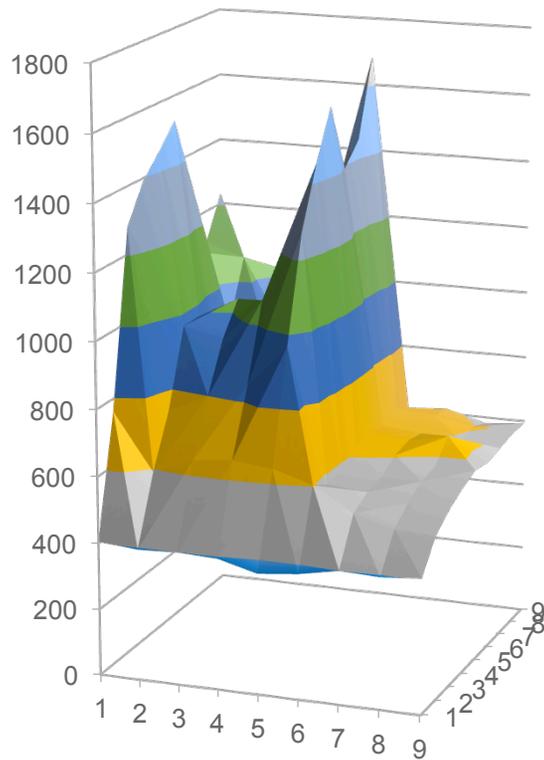


- Performance plummets when matrices no longer fit in cache

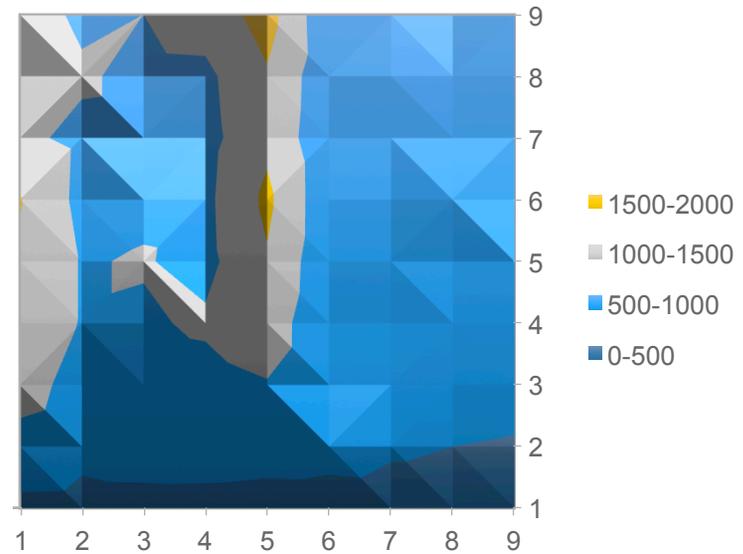
# Blocking for cache helps

- do  $jj=1, n, stride_j$   
do  $ii=1, n, stride_i$   
do  $j=jj, \min(n, jj+stride_j-1)$   
do  $i=ii, \min(n, ii+stride_i-1)$   
 $b(i,j) = a(j,i)$
- Good choices of  $stride_i$  and  $stride_j$  can improve performance by a factor of 5 or more
- But what are the choices of  $stride_i$  and  $stride_j$ ?

# Results: Blue Waters O1



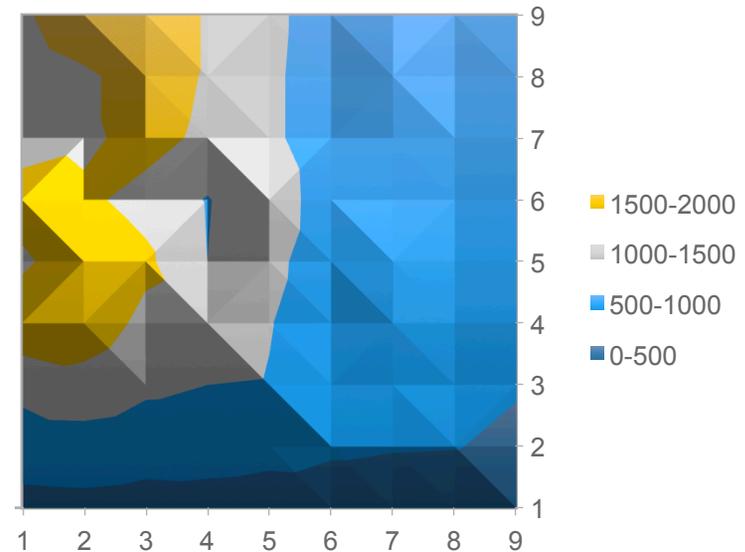
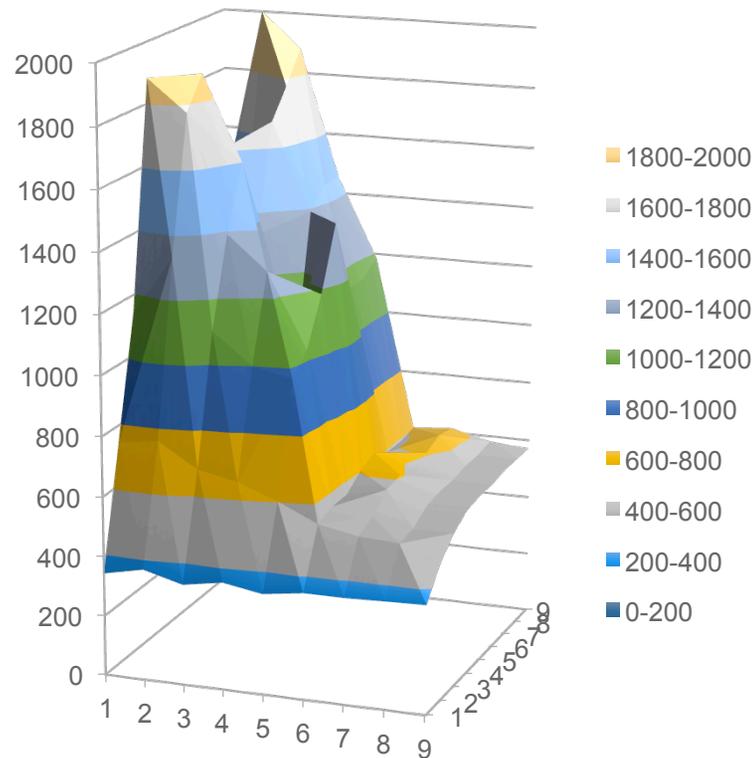
- 1600-1800
- 1400-1600
- 1200-1400
- 1000-1200
- 800-1000
- 600-800
- 400-600
- 200-400
- 0-200



- 1500-2000
- 1000-1500
- 500-1000
- 0-500

# Results: Blue Waters O3

Simple, unblocked code compiled with O3 – 709MB/s



# An Example: Stencil Code from a Real Application

- Stencil for CFD code
- Supports 2D and 3D
- Supports different stencil widths
- Matches computational scientists' view of the mathematics

```
! GICE block=StrnRate
do i = 1,ND
  do k = 1,ND ! diagonal components first
    do ii = 1, Nc
      StrnRt(ii,i) = StrnRt(ii,i) + k
      NT1(ii,i+k*ND-2) = VelGrad1st(ii,i+k*ND-2)
    end do
  end do ! k
  do j = i+1,ND ! upper-half part of strain-rate tensor due to symmetry
    do k = 1,ND
      do ii = 1, Nc
        StrnRt(ii,i+j*ND-2) = StrnRt(ii,i+j*ND-2) + k
        NT1(ii,k+j*ND-2) = VelGrad1st(ii,i+k*ND-2) + k
        NT1(ii,k+i*ND-2) = VelGrad1st(ii,j+k*ND-2)
      end do
    end do ! k
  end do ! j
  do ii = 1, Nc
    StrnSt(ii,i+j*ND-2) = 0.5_rfract * StrnRt(ii,i+j*ND-2)
  end do
end do ! i
do k = 1,size(StrnRt,2)
  do ii = 1, Nc
    StrnRt(ii,k) = JAC(ii) * StrnRt(ii,k)
  end do
end do ! k
! GICE endblock
```

# Another Version of the Same Code

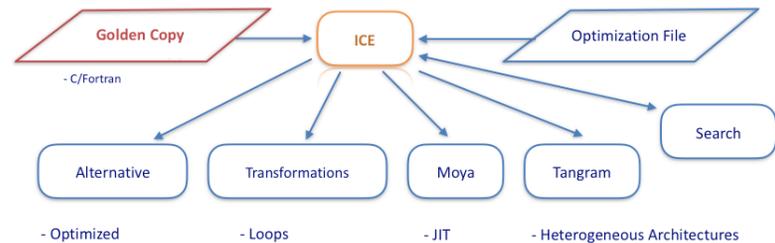
- This version is 4X as fast as the simpler, easier to read code
- Less general code (subset to stencil, problem dimension)
- Same algorithm, data structure, and operations, but transformed to aid compiler in generating fast (and vectorized) code

```
if( ND == 2 ) then
  do ii = 1, Nc
    ! diagonal components first
    StrnRt(ii,1) = JAC(ii) + (
      NTi(ii,1) = VelGradIst(ii,1)
      + NTi(ii,3) = VelGradIst(ii,3) )
    StrnRt(ii,2) = JAC(ii) + (
      NTi(ii,2) = VelGradIst(ii,2)
      + NTi(ii,4) = VelGradIst(ii,4) )
    StrnRt(ii,3) = JAC(ii) + 0.5_rfrreal * (
      NTi(ii,3) = VelGradIst(ii,1)
      + NTi(ii,1) = VelGradIst(ii,3)
      + NTi(ii,4) = VelGradIst(ii,3)
      + NTi(ii,2) = VelGradIst(ii,4) )
  end do
else if( ND == 3 ) then
  do ii = 1, Nc
    ! diagonal components first
    StrnRt(ii,1) = JAC(ii) + (
      NTi(ii,1) = VelGradIst(ii,1)
      + NTi(ii,2) = VelGradIst(ii,4)
      + NTi(ii,3) = VelGradIst(ii,7) )
    StrnRt(ii,4) = JAC(ii) + 0.6_rfrreal * (
      NTi(ii,4) = VelGradIst(ii,1)
      + NTi(ii,1) = VelGradIst(ii,3)
      + NTi(ii,5) = VelGradIst(ii,4)
      + NTi(ii,2) = VelGradIst(ii,5)
      + NTi(ii,6) = VelGradIst(ii,7)
      + NTi(ii,3) = VelGradIst(ii,8) )
  end do
end if

do ii = 1, Nc
  StrnRt(ii,2) = JAC(ii) + (
    NTi(ii,4) = VelGradIst(ii,2)
    + NTi(ii,5) = VelGradIst(ii,5)
    + NTi(ii,6) = VelGradIst(ii,8) )
  StrnRt(ii,6) = JAC(ii) + 0.5_rfrreal * (
    NTi(ii,7) = VelGradIst(ii,2)
    + NTi(ii,4) = VelGradIst(ii,3)
    + NTi(ii,8) = VelGradIst(ii,5)
    + NTi(ii,5) = VelGradIst(ii,6)
    + NTi(ii,9) = VelGradIst(ii,8)
    + NTi(ii,6) = VelGradIst(ii,9) )
  end do
do ii = 1, Nc
  StrnRt(ii,3) = JAC(ii) + (
    NTi(ii,7) = VelGradIst(ii,3)
    + NTi(ii,8) = VelGradIst(ii,6)
    + NTi(ii,9) = VelGradIst(ii,9) )
  StrnRt(ii,5) = JAC(ii) + 0.5_rfrreal * (
    NTi(ii,7) = VelGradIst(ii,1)
    + NTi(ii,1) = VelGradIst(ii,3)
    + NTi(ii,8) = VelGradIst(ii,4)
    + NTi(ii,2) = VelGradIst(ii,6)
    + NTi(ii,9) = VelGradIst(ii,7)
    + NTi(ii,3) = VelGradIst(ii,9) )
  end do
end if
```

# Illinois Coding Environment (ICE)

- One pragmatic approach
- Assumptions
  - Fast code requires some expert intervention
  - Can't all be done at compile time
  - Original code (in standard language) is maintained as reference
  - Can add information about computation to code
- Center for Exascale Simulation of Plasma-Coupled Combustion
  - <http://xpacc.illinois.edu>



- Approach

- Annotations provide additional descriptive information
  - Block name, expected loop sizes, etc.
- Source-to-source transformations used to create code for compiler
  - Exploit tool ecosystem – interface to existing tools
  - Original “Golden Copy” used for development, correctness checks
- Database used to manage platform-specific versions; detect changes that invalidate transformed versions

# Example: Dense Matrix Multiply

## ▶ Matrix Multiplication

```
#pragma @ICE loop=matmul
  for (i = 0; i < n; i++)
    for (j = 0; j < n; j++)
      for (k = 0; k < n; k++)
        mC[i][j] += mA[i][k] * mB[k][j];
#pragma @ICE endloop
```

+

```
---
#Compilation command before tests
buildcmd: make realclean; make CC={compiler} COPT={params}

search:
  tool: opentuner
  time-limit: 30000
  variants-limit: 1000

buildoptions:
  gcc:
    params: {'-O': {'default': 3, 'min': 0, 'max': 3}}

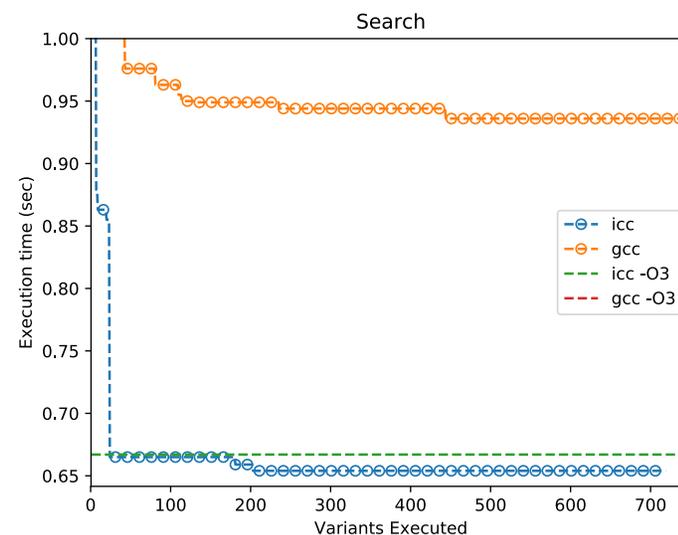
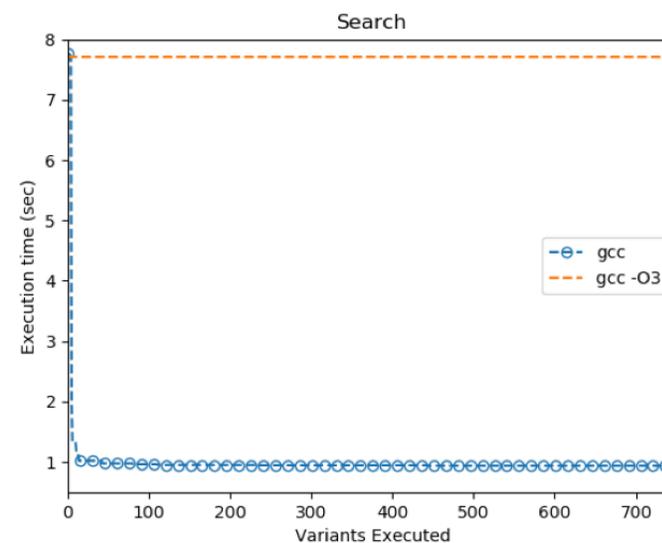
#Command call for each test
runcmd: ./mmc

tuning: on

matmul:
  rose_uiuc:
    - stripmine+:
      loop: 3
      factor: 2..36
    - stripmine+:
      loop: 2
      factor: 2..48
    - interchange+:
      order: 1,3,0,2,4
    - unroll*:
      loop: 5
      factor: 2..24
...
```

# Performance Results

- Dense matrix-matrix multiply
  - 302,680 total variants
  - Subset evaluated (based on results-so-far)
  - 8.2x speedup over gcc compiler with optimization
  - Small but consistent speedup over icc -O3
- Different parameters can be selected/remembered for each platform
  - Within the constraints of the performance parameters considered



# Stencil 3D

```
#pragma @ICE loop=stencil
for(i = 1; i < x-1; i++) {
  for(j = 1; j < y-1; j++) {
    for(k = 1; k < z-1; k++) {
      B[i][j][k] = C0 * A[i][j][k] + C1 * (
        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]);
    }
  }
}
#pragma @ICE endloop
```

+

```
---
#Built command before compilation
prebuilddcmd:

#Compilation command before tests
builddcmd:
    make realclean; make CC={compiler} COPT={params}

buildoptions:
  gcc:
    params: {'-O': {'default': 3, 'min': 0, 'max': 3}}
  icc:
    params: {'-O': {'default': 3, 'min': 0, 'max': 3}}

#Command call for each test
runcmd: ./sten3d 1024 20

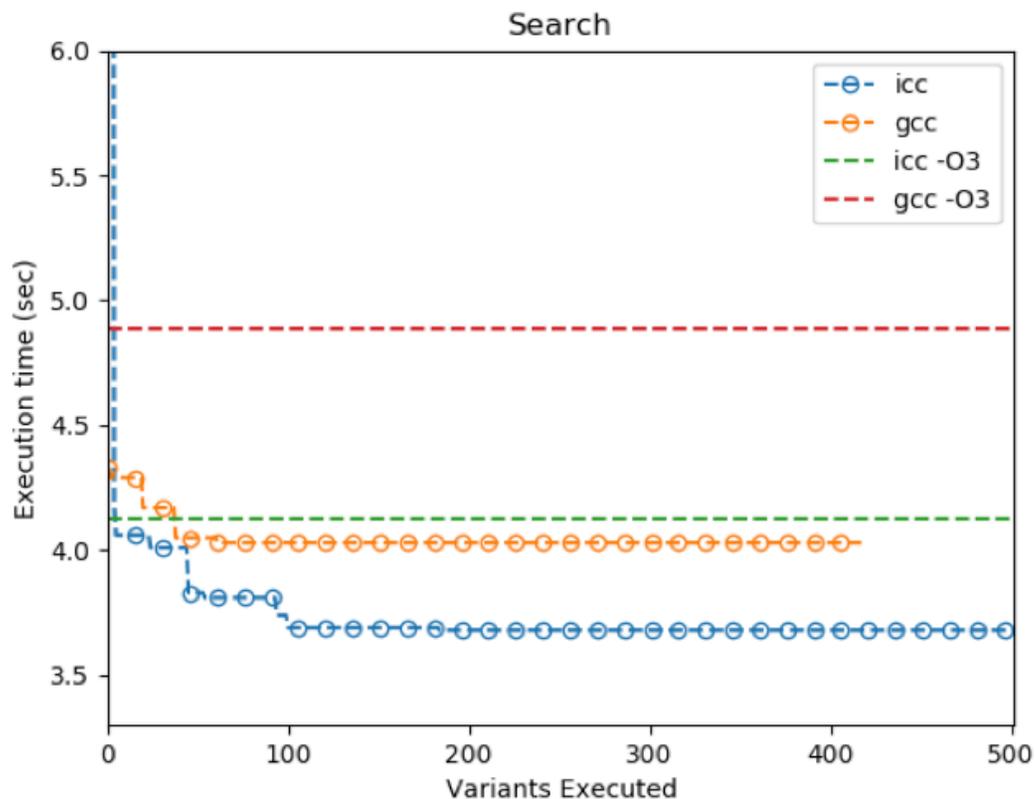
tuning: on

stencil:
  rose_uiuc:
    - stripmine+:
      loop: 4
      factor: 16..1024
      type: poweroftwo
    - stripmine+:
      loop: 3
      factor: 16..1024
      type: poweroftwo
    - stripmine+:
      loop: 2
      factor: 16..1024
      type: poweroftwo
    - interchange+:
      order: 0,1,3,5,2,4,6
```



# Performance Results

- 3-D Stencil
  - 11,664 variants
  - Max 12.6 sec
  - Min 3.68 sec
  - Speedup over simple code
    - icc: 1.12x
    - gcc: 1.21x



# Other Dangers

- How do we know that the performance portable code is correct?
  - Or even if it will compute the same result as the original code
    - And what is “the same result”?
- It is *not enough* to prove that any code transformations are correct
  - MPICH used to test whether the compiler returned the same result in a and c for these two statements:
    - `a = joe->array[OFF+b+1];`  
`c = joe->array[OFF+1+b];`
  - Because one major vendor compiler got this *wrong*.
- And you still need to prove that the hardware implements all of the operations correctly
  - And vectorization is already likely to produce results that are not bitwise identical to the non-vector version (which might depend on how data is aligned at runtime)
- Question: How do you test that the performance portable code is computing what is intended?
- Proving code transformations correct is *necessary* but not *sufficient*

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# So What Is Performance Portability?

- Rather than define whether a code is (or is not) performance portable, look at the goals
  - Make it easier for end users to run an application code effectively on different systems
    - for some set of systems – not necessarily every possible system
    - May focus on the workflow or the I/O performance, rather than any single code
  - Make it easier for developers to write, tune, and maintain an application for multiple systems
    - Allows a tradeoff between one code and several, based on what's *easier*

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# Summary

- Don't underestimate the difficulty
  - I don't believe "strong" performance portability is possible
- Don't give up
  - There is a lot that can be done to support users and improve performance resilience
- Accept different approaches
  - Different communities, expectations, goals
- Be precise about your goal and accomplishment
  - Let this be a No Hype zone