An introduction to OpenMP

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Before We Start

• The OpenMP web site is a good source of information:

openmp.org

- tutorials and examples from beginner to advanced
- the standard (which is easy to understand for a standard)
- quick reference guides

A good collection of simple reference examples

users.abo.fi/mats/PP2012/examples/OpenMP/

Multicore

The Free Lunch

- For a long time high-performance computing had a "free-lunch"
 - The density of transistors in chips increased, decreasing the size of integrated circuits
 - same number of transistors with less power
 - more transistors to add functionality
 - The clock speeds steadily rose, increasing the number of operations per second (from MHz to GHz)
- But the free lunch has been over for a few years now
 - We are reaching the limitations of transistor density
 - Increasing clock frequency requires too much power

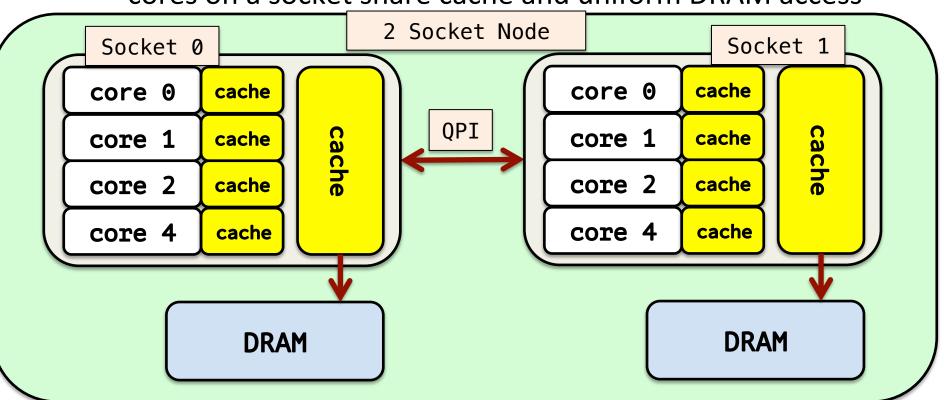
we used to focus on floating point operations per second now we also think about floating point operations per Watt

The Solution: Multicore

- The number of transistors is still increasing
 - Sandy Bridge 28 nm, Ivy Bridge /Haswell 22 nm
 - Projected hard limit of 5-7 nm
- This has lead to three trends
 - more cores
 - the Sandy Bridge processor on Daint has 8 cores
 - Xeon phi KNL will have 72 cores
 - reducing clock speed
 - simplify/specialize cores
 - An extreme example of this is GPUs, which have in the order of 100/1000s of cores specialized for tasks common in graphics

Multicore HPC Nodes

- A CPU-based node can have multiple sockets, each with multiple cores.
 - cores on a socket share cache and uniform DRAM access

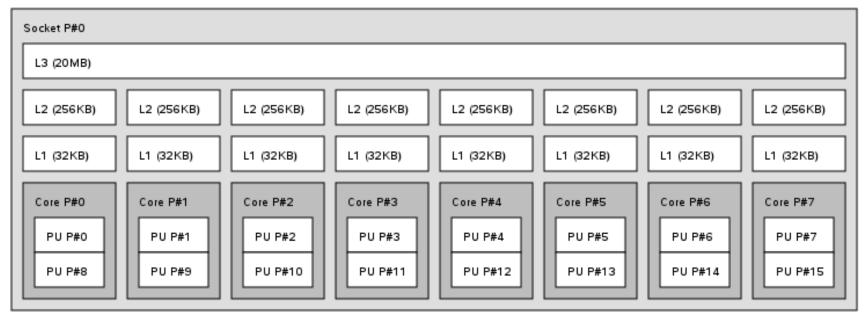




Piz Daint: using Istopo

 the lstopo command can be used to visualize node topology

Machine (31 GB)



The Flat MPI Model

- MPI is the dominant parallelization model in HPC
 - The problem being solved is broken into chunks, one chunk per MPI process
 - Processes communicate via message passing
- The flat MPI model was well-suited to few cores per socket/node
 - Each MPI process sees one core with its cache and memory.
 - It is a logical abstraction for the underlying hardware
- You will learn about MPI later in this course

The Hybrid MPI-OpenMP Model

- The flat MPI model assigns one process per core
 - for 8/16/32/64 ranks per multi-core node
 - does not scale to many nodes: the amount of data passed around in messages increases as number of ranks increases
 - to take advantage of shared cache and DRAM on a socket, why not use threads on the socket/node, and pass messages between sockets/nodes?
- The hybrid MPI-OpenMP model has light-weight threads that share on-node memory.

This course is about OpenMP, however you could use other threading technologies like pthreads, Cilk++, Intel
Threading Building Blocks or C++11 threads.

NUMA

- On a multi-socket node each socket has its own DRAM memory
- However the operating system presents all memory as one unified memory space
 - threads can read and write to memory that is local to other sockets.
 - non-local accesses (to local memory on another socket)
 are slower than local accesses
- This memory architecture is called Non-Uniform Memory Access (NUMA)
 - "NUMA effects" arise when threads frequently access memory on a different NUMA domain.

Avoiding NUMA Effects

- Avoid them: hybrid MPI-OpenMP with at least one rank per NUMA node
 - threads in an MPI rank all run on the same NUMA node
 - explicit message passing via MPI between NUMA nodes
- Use first-touch memory allocation
 - by default the Linux kernel assigns memory pages to the NUMA node that first "touches" the memory
 - it is not the allocation of memory, but the first write to the memory that is important
 - e.g. have the thread that will be accessing a part of memory initialize
 - memory allocation methods that initialize memory values are to be avoided, e.g. calloc and std::allocator<>

NUMA First Touch Example

```
const int N = 10000000000;
double *x = (double*)malloc(N*sizeof(double));
double *x = (double*)malloc(N*sizeof(double));
double *y = (double*)malloc(N*sizeof(double));
// initialize memory as it will be used
#pragma omp parallel for
for(i=0; i<n; ++i) {
 x[i] = ...;
 z[i] = \ldots;
#pragma omp parallel for
for(i=0; i<n; ++i)
  z[i] = x[i] + y[i];
```

Affinity

- CPU affinity is the pinning of a process or thread to a core
 - If the operating system interrupts the thread, it doesn't migrate it to another core
 - For most HPC scenarios where only one application is running on a node, these interruptions are short
 - Some multi-core architectures (e.g. IBM BG/Q) require multiple threads/core to achieve peak performance
 - On x86 the best results are obtained with 1 or 2 threads per core
- Memory affinity is allocation of memory as closely as possible to the cores that will be accessing it
 - achieved by first touch or MPI between NUMA nodes

Controlling Thread Affinity on Daint

- On Cray systems the aprun command used to launch jobs provides flags for setting thread affinity
 - see the man page for all options > man aprun
- The best configuration can be system and application dependent
- letting the runtime schedule threads is useful when using libraries that create additional threads

OpenMP

What is OpenMP?

Open Multi
Processing

- The OpenMP standard is an API with compiler directives, a run time library, and environment variables for writing parallel shared memory applications in C, C++ and Fortran.
 - supported by compiler vendors in HPC, including GNU,
 Intel, Cray and PGI compilers. Also Clang and MSVC.
- OpenMP compilers allow programmers to tell the compiler where and how to parallelize key loops and tasks with directives
 - The user does not have the complication of explicitly managing threads, as would be the case with a low level threading library like pthreads.

Goals of OpenMP

- Standardization
- Lean
 - concise and simple set of directives
 - possible to get good speedup with a handful of directives
 - but each new release gets more complicated (now at 4.0)
- Ease of use
 - you can incrementally add it to code without major changes
 - in practice refactoring is required to get performance
- Portability
 - Supported by range of compilers and platforms

Should I use OpenMP?

- OpenMP + MPI often doesn't perform as well as flat MPI on multicore
 - at least not without significant developer effort
- But for many-core architectures like Xeon Phi flat MPI does not scale
 - lightweight threading is required
- OpenMP is the defacto standard for Fortran, and path of least resistance existing codes
- Other options exist for C and C++
 - look into them if developing new code, or if you have a lot of task-based parallelism

OpenMP Compiler Directives

 In C and C++ parallel regions are scopes that are marked with #pragma omp parallel

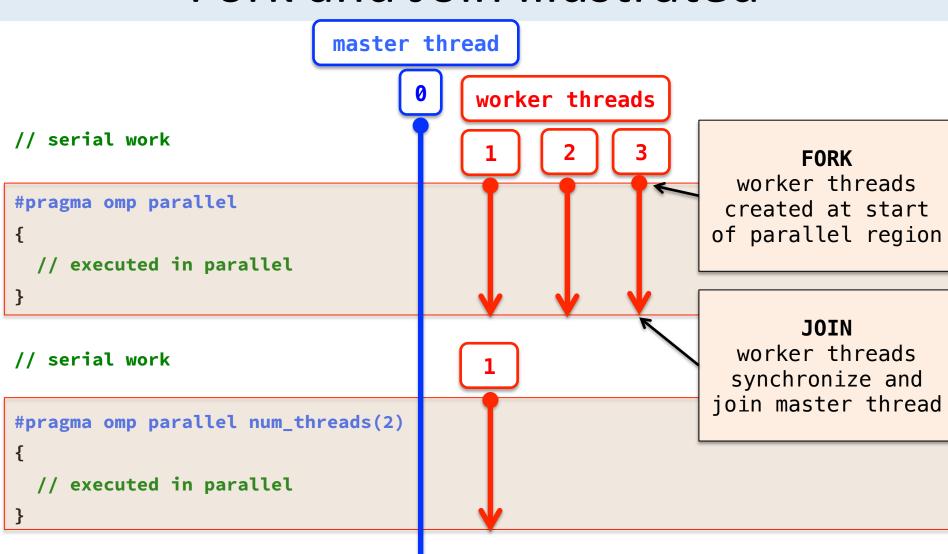
 In Fortran parallel regions are indicated with specially-formatted comments

```
!$omp parallel
! executed in parallel
!$omp end parallel 
### must have comment at start and end of parallel region
```

Fork and Join Model

- OpenMP uses a fork and join model for threading
- The application starts with a master thread
 - FORK: a team of parallel worker threads is started at the beginning of each parallel block
 - the block is executed in parallel by each thread
 - JOIN: the worker threads are synchronized at the end of the parallel block and join with the master thread.
- Threads are numbered 0:N-1, where N is the total number of threads
- The master thread is always numbered 0.

Fork and Join Illustrated





Compiling OpenMP

- Most compilers require a flag to enable OpenMP compilation
 - without a flag the #pragma or !\$omp directives are ignored by the compiler and a serial application is created
- compilers that don't understand OpenMP will simply ignore the directives (no portability problems).

```
Cray : on by default for -01 and greater, disable with -h noomp
```

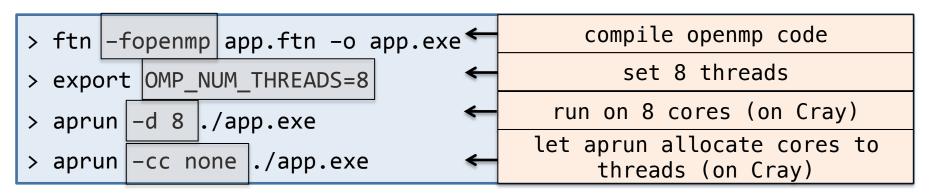
Intel : off by default, enable with -openmp

GNU : off by default, enable with -fopenmp

PGI : off by default, enable with -mp

Running OpenMP applications

- The default number of threads is set with an environment variable OMP NUM THREADS
- There has to be at least one core per thread
 - multiple threads on a single core have to share resources on the core
 - On Cray systems, like Daint, the best approach seems to be to let aprun assign the threads (-cc none)



Exercises: before starting

- The source code for exercises and slides are in the repository
 - to get a copy in your local path

```
> ssh -X username@ela.cscs.ch
```

- > ssh daint
- > cd \$SCRATCH
- > git clone https://github.com/fomics/SummerSchool2015
 - for these examples we will use the gnu compiler, for which a script to set up the environment is provided
- > source setup.sh
- > CC hello_world.cpp -fopenmp
- > ftn hello_world.f90 -fopenmp

we use the GNU compiler because the Cray compiler aggressively replaces simple loops with BLAS calls in the absence of OpenMP directives, which make it difficult to compare OpenMP versions of a code.

let us know if you have a problem

getting the code

Exercise 1: Compiling and running

- Open the test code hello_world.cpp/f90
 - what do you expect the output to be?
- Compile: choose C++ or Fortran version

 > module swap PrgEnv-cray PrgEnv-gnu
 - > CC hello_world.cpp -fopenmp
 - > ftn hello_world.f90 -fopenmp
- Then run

Is the output what you expected? Why?

Exercise 1: Affinity

- go to the openmp/exercises/affinity path
- Compile:
 - > cd SummerSchool2015/openmp/exercises/affinity
 - > make
- Experiment with affinity
 - try different numbers of threads and aprun flags

```
> OMP_NUM_THREADS=1 aprun -cc none ./test.omp
```

- > OMP_NUM_THREADS=8 aprun -cc none ./test.omp
- > OMP_NUM_THREADS=8 aprun -d 8 ./test.omp
- > OMP_NUM_THREADS=8 aprun -d 4 ./test.omp

Runtime Library

- OpenMP has runtime library routines for controlling your application, including
 - int omp_get_thread_num()
 - get id of current thread
 - int omp_get_num_threads()
 - number of threads in current parallel region
 - int omp_get_max_threads()
 - default number of threads in parallel regions: corresponds to OMP_NUM_THREADS environment variable
 - double omp_get_wtime()
 - accurate timing function: returns double
- There are many others, however these are the most commonly used

Runtime Library

 The runtime library requires that the omp header/ module is included

```
use omp_lib
#include <omp.h>
                                          integer :: threads, inside, outside
int threads = omp_get_max_threads();
int outside = omp_get_num_threads();
                                          threads = omp_get_max_threads()
                                          outside = omp_get_num_threads()
int inside;
#pragma omp parallel
                                          !$omp parallel
                                          inside = omp_get_num_threads()
                                          !$omp end parallel
  inside = omp_get_num_threads();
                                          print *, inside, ' in ', outside, ' out ',
                                                    threads, ' max'
printf("%d in, %d out, %d max \n",
       inside, outside, threads);
                   > OMP_NUM_THREADS=8 ./a.out
```

8 in, 1 out, 8 max

Synchronization

- Sometimes you need to synchronize threads inside a parallel region, for example
 - tasks that have to be done by only one thread
 - when multiple threads have to update memory, and we need to ensure that the value in memory is consistent
 - threads have to wait for others to finish before continuing
- OpenMP provides directives that can be used to indicate such regions
- WARNING: synchronization and serial code regions can quickly limit the potential speed up from parallelism.

Synchronization Example

 The intention of the code below is to print a hello world message, similarly to the hello world example earlier:

```
#pragma omp parallel
{
  int tid = omp_get_thread_num();
  printf("hello world from thread %d", tid);
}

> OMP_NUM_THREADS=8 ./a.out
???
```

 each thread gets a private copy of tid, but the output could get messed up because all threads write to stdout at the same time.

master Directive

- The master directive indicates sections that are to be executed only by the master thread
 - the master thread is always thread 0

```
#pragma omp parallel
{
  int tid = omp_get_thread_num();
  #pragma omp master
  printf("hello world from thread %d", tid);
}
> OMP_NUM_THREADS=2 ./a.out
hello world from thread 0
```

only the master thread prints its tid value.

single Directive

- Block will only be executed by the first thread to arrive at the block
 - varies from one run to the next

```
#pragma omp parallel
{
  int tid = omp_get_thread_num();
  #pragma omp single
  printf("hello world from thread %d", tid);
}
> OMP_NUM_THREADS=8 ./a.out
hello world from thread 3
> OMP_NUM_THREADS=8 ./a.out
hello world from thread 6
```

only one thread will print message.

critical Directive

- All threads will execute block, one at a time
 - in order that the threads arrive at the block: varies each time application is run

```
#pragma omp parallel
{
  int tid = omp_get_thread_num();
  #pragma omp critical
  printf("hello world from thread %d", tid);
}
> OMP_NUM_THREADS=3 ./a.out
hello world from thread 1
hello world from thread 0
hello world from thread 2
```

 each thread prints one message, in order of thread arrival at the critical block

barrier Directive

 All threads wait at barrier until all of the threads are at barrier, before continuing

```
#pragma omp parallel
{
  int tid = omp_get_thread_num();
  #pragma omp barrier
  printf("hello world from thread %d", tid);
}
> OMP_NUM_THREADS=3 ./a.out
hello world from thread 0 hello world from thread 3
hello world from thread 2
```

 in this case, this doesn't solve anything: all threads try to write to stdout at the same time when they simultaneously leave the barrier

Exercise 2

- Go back to the hello_world.cpp/f90 example, and add appropriate synchronization directives
 - do you see expected behavior now?
- Look at sum_threads.cpp/f90
 - what was the intended output?
 - run the example and compare the actual output.
 - can you add a synchronization directive to get the expected result?

```
> source setup.sh
> CC sum_threads.cpp -fopenmp
... or
> ftn sum_threads.f90 -fopenmp
> OMP_NUM_THREADS=8 aprun -cc none ./a.out
```

Shared Memory Model

- OpenMP uses a shared memory model
- All threads can read and write to the same memory locations simultaneously
- By default variables are shared, so one copy is used by all threads
- The result of computations where multiple threads attempt to read/write to a variable are undefined
 - see the sum_threads example in the previous exercises for a simple example
 - this is a very common parallel programming bug called a race condition

Variable Scoping

- OpenMP provides clauses that describe how variables should be shared between threads
 - shared: all variables access the same copy of a variable.
 - this is the default behavior
 - WARNING: take care when writing to shared variables
 - private: each thread gets it's own copy of the variable
 - private copy is uninitialized
 - use firstprivate to initialize variable with value from master

```
int tid;
const int num_threads = omp_get_num_threads();
#pragma omp parallel shared(num_threads) private(tid)

{
   tid = omp_get_thread_num();
   #pragma omp critical
   printf("hello world from thread %d of %d", tid, num_threads);
}
```

Private Variables in C99/C++

- Variables that are declared inside a parallel region are private by default
 - this isn't possible in Fortran because variables must be declared at the top of each subroutine/program/function

```
int tid;
#pragma omp parallel private(tid)
{
  tid = omp_get_thread_num();
  // ... use tid for local computation
}
using private clause
```

```
#pragma omp parallel
    best practice: declare variable in scope

{
    int tid = omp_get_thread_num();
    // ... use tid for local computation
}
```

Exercises 3

 Now we finally have all the tools needed to fix hello_world.cpp/f90

Work Sharing: for/do loops

- A common target for for parallelization is loops without loop carried dependencies
 - for example, adding two vectors:

- we could attempt to parallelize this using the techniques that we have learnt so far...

for/do loops the hard way

```
double *x, *y, *z;
int n;
...
for(int i=0; i<n; ++i) {
   z[i] = x[i] + y[i];
}
   serial</pre>
```

What a mess!
And error-prone
too: does this
approach still work
if n<num_threads?

```
double *x, *y, *z;
                           calculate loop
int n;
                           bounds for this
                           thread's chunk
#pragma omp parallel
 int tid = omp_get_thread_num();
  int num_threads = omp_get_num_threads();
  int work = n/num_threads;
  int s = tid*work;
  int e = (tid==num_threads-1) ? n : s+work;
  for(int i=s; i<e; ++i)</pre>
   z[i] = x[i] + y[i];
                                      parallel
```

parallel for

OpenMP provides a directive for for/do loops

```
loop index
double *x, *y, *z;
                                           real(kind=8) :: x(:), y(:), z(:)
                         variable i is
                                                   :: i, n
int n, i;
                                           integer
                           private by
                             default
#pragma omp parallel
                                           !$omp parallel
                                           !$omp do
 #pragma omp for
                                           d i=1,n
 for(i=0; i<n; ++i) {
                                             z(i) = x(i) + y(i)
   z[i] = x[i] + y[i];
                                           end do
                                            !$omp end do
                                 C++
                                                                                 Fortran
                                            !$omp end parallel
```

- Compiler handles loop bounds for you
- there is a compact single-line directive:

```
#pragma omp parallel for
for(i=0; i<n; ++i) {
    z[i] = x[i] + y[i];
}</pre>
```

Example: Vector Normalize

- Open vector_normalize.cpp/f90
 - write a parallel version of the function/subroutine normalize_vector()
 - find the norm of the vector v
 - scale the vector by the norm to give it length=1
- 2. Open dot.cpp/f90
 - this code finds the dot product of two vectors
 - add OpenMP directives to parallelize the code

Reductions

 Reduction operations reduce a set of values to a single value according to an operation op

```
a = initial value
for i = 1,n
a = a op expr
```

```
e.g. C++ sum reduction

double sum = 0.0, v[n];

for(int i=0; i<n; ++i)
   sum = sum + v[i];</pre>
```

- OpenMP provides a clause of the form reduction(op:list) for performing reductions
 - a is a scalar variable in list
 - expr is a scalar expression that does not reference a
 - only certain expressions allowed
 e.g. (+,-,*,/,binary ops)

```
double sum = 0.0, v[n];
#pragma omp parallel for
#pragma omp reduction(+:sum)
for(int i=0; i<n; ++i)
  sum = sum + v[i];</pre>
```

Exercise

- Revisit the dot.cpp/f90 exercise, and rewrite it to use a reduction.
 - use test.sh to see how it scales from 1 to 8 threads
 - try arrays of length 10'000, 100'000 and 1'000'000
 - how does this affect scaling?
- We can apply everything that we have learnt to the example pi.cpp/f90, which computes pi using the trapezoidal rule
 - use test.sh to test its scaling from 1 to 8 threads
 - is it's scaling better or worse? why?

Nested Loops

- If you apply directives nested loops, the outer loop will be parallelized.
- the collapse directive will merge the loops and parallelize both

```
double sum = 0.0, v[n];
#pragma omp parallel for
for(int j=0; j<n; ++j) {
  for(int i=0; i<m; ++i)
    A[j][i] = f(B[j][i]);
}
  n parallel work items</pre>
```

```
double sum = 0.0, v[n];

#pragma omp parallel for collapse(2)
for(int j=0; j<n; ++j) {
  for(int i=0; i<m; ++i)
    A[j][i] = f(B[j][i]);
}
    n*m parallel work items</pre>
```

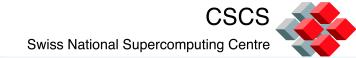
 use if n is small relative to OMP_NUM_THREADS and/or the amount of work in f is significant

Limitations to Parallel Speedup

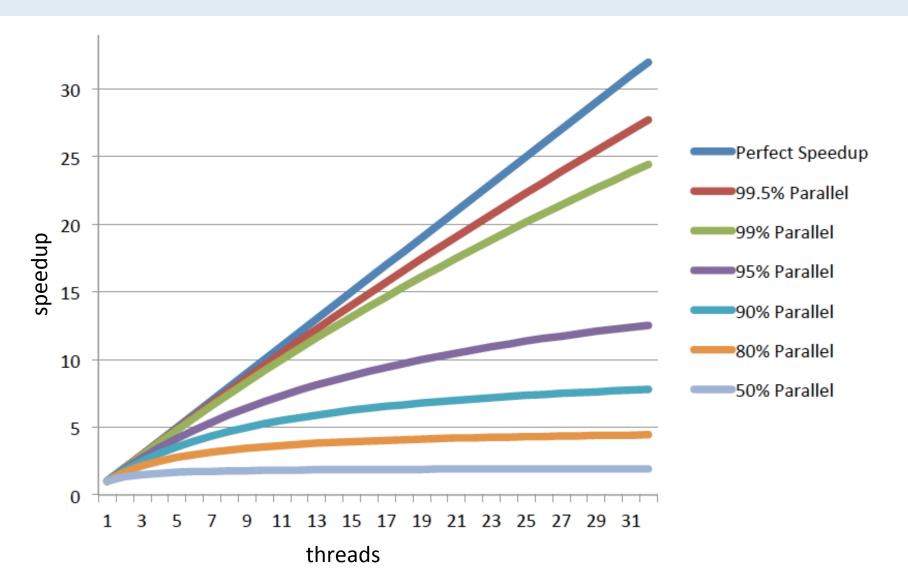
Limitations To Parallel Speedup

- The amount of speedup that you will get from OpenMP is dependent on many factors including
 - having enough work to keep all threads busy
 - the ratio of sequential to parallel work
- Amdahl's Law is used to define the maximum possible speedup when only parts of a code can be parallelized
 - T(n) = T(1) * (B+(1-B)/n)
 - i.e. T(∞)=T(1)*B
 - where T(n) is time to solution for n threads, and B is the fraction of the algorithm that is strictly serial





Amdahl Illustrated



Amdahl's Law

- Our application would ideally scale to many threads
- But just 1% serial code in an algorithm means that it won't scale more than 100 times: no matter how many threads/nodes we use.
- While adding OpenMP directives to an existing sequential code is easy, the modifications required to get really good scaling might require significant restructuring
 - there is no free lunch!

OpenMP 4.0 and Accelerators

- OpenMP 4.0 adds directives for using accelerators
 - Including GPUs and Intel Xeon Phi (MIC)
 - GCC 5.0 and the Intel compilers support of Xeon Phi offload
 - Currently there is no OpenMP 4.0 GPU offload
- Similar concept to the OpenACC directives that you will see for GPUs next week