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Overview

29. May 2017

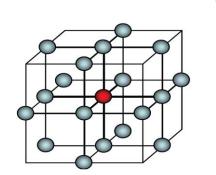
- This worked example leads you through accelerating a simple application
 - a simple application is easy to understand
 - but it shows all the steps you would use for a more complicated code
- Beside the theoretical background, we will implement the OpenMP directives and compare results.

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The Himeno Benchmark

- 3D Poisson equation solver
 - Iterative loop evaluating 19-point stencil
 - Memory intensive, memory bandwidth bound
- Fortran and C implementations available
- We look at the scalar version for simplicity
- Code characteristics
 - Around 230 lines of Fortran or C
 - Arrays statically allocated
 - problem size fixed at compile time



Why use such a simple code?

- Understanding a code structure is crucial if we are to successfully OpenMP an application
 - i.e. one that runs faster node-for-node
 - not just full accelerator vs. single CPU core
- There are two key things to understand about the code:
 - How is data passed through the calltree?
 - Where are the hotspots?
- Answering these questions for a large application is hard
 - There are tools to help
 - we will discuss some of them later in the tutorial
 - With a simple code, we can do all of this just by code inspection

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The key questions in detail



- How is data passed through the calltree?
 - CPUs and accelerators have separate memory spaces
 - The PCIe link between them is relatively slow
 - Unnecessary data transfers will wipe out any performance gains
 - A successful OpenMP/OpenACC port will keep data resident on the accelerator
- Where are the hotspots?
 - The OpenMP/OpenACC programming model is aimed at loop-based codes
 - Which loopnests dominate the runtime?
 - Are they suitable for an accelerator?
 - What are the min/average/max tripcounts?
- Minimising data movements will probably require acceleration of many more (and possibly all) loopnests
 - Not just the hotspots
 - any loopnest that processes arrays that we want accelerator-resident
 - But we have to start somewhere

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First stages to accelerating an application



Understand and characterise the application

Profiling tools, code inspection, speaking to developers if you can

1. Introduce first OpenMP kernels

2. Introduce data regions in subprograms

- reduce unnecessary data movements
- will probably require more OpenMP kernels

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Next stages to accelerating an application

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- 3. Move up the calltree, adding higher-level data regions
 - ideally, port entire application so data arrays live entirely on the GPU
 - otherwise, minimise traffic between CPU and GPU
 - This will give the single biggest performance gain
- 4. Only now think about performance tuning for kernels
 - First correct any obviously inefficient scheduling on the GPU
 - This will give some good performance improvements
 - Optionally, experiment with OpenMP tuning clauses
 - You may gain some final additional performance from this
- And remember Amdahl's law...

We will treat in another session

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Initial run

Reference Understand the application

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Run the initial CPU version of the code

CRA

- First compile and run the code "as is"
 - on the CPU, using one core
 - including a performance profile
- compiled with Cray Compilation Environment (CCE):
 - \$> module switch / load ...
 - \$> module load perftools-base
 - \$> module load perftools-lite-loops
 - \$> make VERSION=00
 - \$> sbatch submit.wlm

The application output:

The runtime

Time (secs): 3.4616304183145985

 The performance (from the runtime, in MFLOPS)

MFLOPS : 3960.7098225914551

 A residual value from the solver (checksum)

Gosa : 0.137835972438427479E-02

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Is the code still correct?

- Most important thing is that the code is correct:
 - Make sure you check the residual (Gosa)
 - N.B. will never get bitwise reproducibility between CPU and GPU architectures
 - different compilers will also give different results
- Advice: make sure the code has checksums, residuals etc. to check for correctness.
 - even if code is single precision, try to use double precision for checking.
 - globally or at least for global sums and other reduction variables

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10

Profile of the initial CPU version of the code



- most expensive computational computation
 - In loop at line 232

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when suitable, accelerate this operation

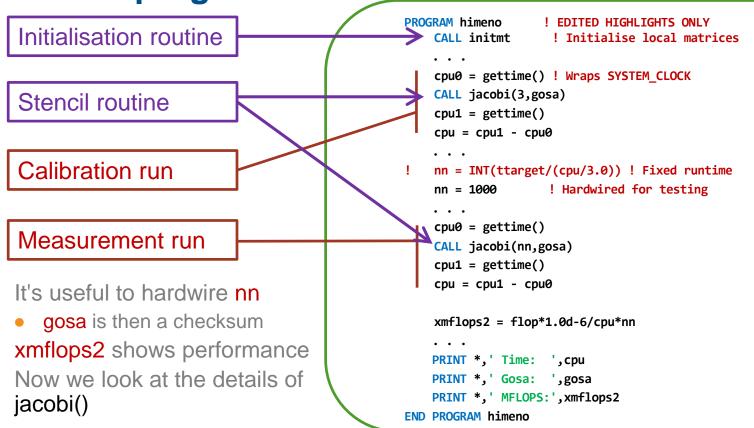
```
$> cat himeno v00.log
         Inclusive and Exclusive Time in Loops (from -hprofile generate)
                      Time
                               Loop Hit
                                         Loop
                                                           Loop | Function \ \ . LOOP[.]
  Loop
            Loop
                                                   Loop
                                          Trips | Trips |
  Incl
            Incl
                      (Loop
                                                          Trips
 Time% |
            Time
                      Adj.)
                                            Avg |
                                                    Min |
                                                            Max
                                                             100 | jacobi .LOOP.1.li.263
  95.9%
          3.576932
                    0.000014
                                            51.5
                                           126.0
                                                             126 | jacobi .LOOP.2.li.270
 82.6%
          3.080879
                    0.000533
                                     103
                                                     126
  82.6%
         3.080346
                    0.049932
                                  12,978
                                           126.0
                                                     126
                                                             126 | jacobi .LOOP.3.li.271
```

Let's have a closer look to the code

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Himeno program structure



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Structure of the jacobi routine

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- Iteration loop
 - must be sequential!
- Apply stencil to p
 - create temporary wrk2
 - residual gosa computed
- Update array p
 - from wrk2
 - can be parallelised
 - outer halo unchanged

```
SUBROUTINE jacobi(nn,gosa)
iter_lp: DO loop = 1,nn
 compute stencil: wrk2, gosa from p
     <described on next slide>
! copy back wrk2 into p
      DO k = 2, kmax-1
         DO j = 2,jmax-1
            DO i = 2, imax-1
             \gg (i,j,k) = wrk2(i,j,k)
            ENDDO
         ENDDO
      ENDDO
   ENDDO iter lp
END SUBROUTINE jacobi
```

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The Jacobi computational kernel

- The stencil is applied to pressure array p
 - 19-point stencil
- Updated pressure values are saved to temporary array wrk2
- Residual value gosa is computed
- This loopnest dominates runtime
 - Can be computed in parallel
 - gosa is reduction variable

```
DO k = 2, kmax-1
 DO j = 2, jmax-1
 DO i = 2,imax-1
   s0=a(i,j,k,1)*p(i+1,j,k) &
    +a(i,j,k,2)*p(i, j+1,k) &
    +a(i,j,k,3)*p(i, j, k+1) &
    +b(i,j,k,1)*(p(i+1,j+1,k)-p(i+1,j-1,k) &
                -p(i-1,j+1,k)+p(i-1,j-1,k)) &
    +b(i,j,k,2)*(p(i, j+1,k+1)-p(i, j-1,k+1) &
                -p(i, j+1,k-1)+p(i, j-1,k-1)) &
    +b(i,j,k,3)*(p(i+1,j,k+1)-p(i-1,j,k+1) &
                -p(i+1,j, k-1)+p(i-1,j, k-1)) &
    +c(i,j,k,1)*p(i-1,j,k) &
    +c(i,j,k,2)*p(i, j-1,k) &
    +c(i,j,k,3)*p(i, j, k-1) &
    + wrk1(i,j,k)
   ss = (s0*a(i,j,k,4)-p(i,j,k)) * bnd(i,j,k)
   gosa = gosa + ss*ss
  wrk2(i,j,k) = p(i,j,k) + omega*ss
  ENDDO
 ENDDO
ENDDO
```

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Compiler feedback



Compiler feedback is extremely important

- Did the compiler recognise the accelerator directives?
 - A good sanity check
- How will the compiler move data?
 - Only use data clauses if the compiler is over-cautious on the map(*)
 - Or you want to declare an array to be scratch (create clause)
 - The first main code optimisation is removing unnecessary data movements
- How will the compiler schedule loop iterations across GPU threads?
 - Did it parallelise the loopnests?
 - Did it schedule the loops sensibly?
 - The other main optimisation is correcting obviously-poor loop scheduling

Generate listing files

CCE:

Produces commentary files <stem>.lst

PGI:

Feedback to STDERR

- Compiler teams work very hard to make feedback useful
- advice: use it, it's free!
 (i.e. no impact on performance)

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Compiler listings



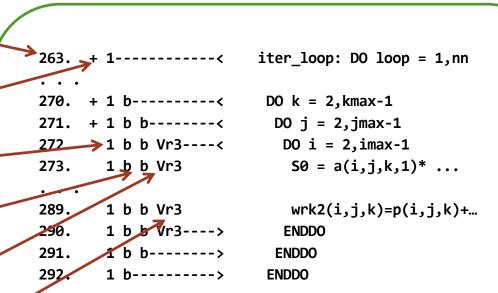
source line numbers

+ = additional message below

Numbers denote serial loops

V = vectorized

r = unrolled



306. 1-----> ENDDO iter_loop

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First kernel

Now we know where, so we implement it and compare.

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Versioning

- Create a new version
 - Thus we can compare
 - We can go back, when we messed up
- Here we simply create a copy of the code
- \$> cp himeno_*_v00.* himeno_*_v01.*

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Step 1: a first OpenMP device kernel

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- Start with most expensive
 - apply teams distribute
- reduction clause
 - Necessary for correct gathering of data
- private clause (optional here)
 - By default:
 - loop variables private (i, j, k)
 - scalar variables are private (s0,ss)
 - Note that private arrays
 - <u>always</u> need <u>private</u> clause

```
qosa1 = 0d0
!$omp target teams distribute &
!$omp& reduction(+:gosa1) &
!$omp& private(i,j,k,s0,ss)
DO k = 2, kmax-1
DO j = 2, jmax-1
  DO i = 2,imax-1
   s0 = a(i,j,k,1) * p(i+1,j, k) &
     <etc...>
   ss = (s0*a(i,j,k,4)-p(i,j,k))* &
        bnd(i,j,k)
   gosa1 = gosa1 + ss*ss
   wrk2(i,j,k) = p(i,j,k) + omega*ss
  ENDDO
 ENDDO
ENDDO
!$omp end target teams distribute
```

Compiler listings: vi himeno_*_v01.lst



Check what the compiler did

```
G = accelerated
```

```
g = partioned
```

```
b = blocked loops
```

```
269. + 1 G------ !$omp target teams distribute reduc...
270. 1 \frac{1}{6} g----- D0 k = 2, kmax-1
271. + 1 \frac{6}{9} b----- D0 j = 2, jmax-1
272.
    1 G g b gb---< D0 i = 2,imax-1
273.
    ___5G g b gb
                       S0 = a(i,j,k,1)*p(i+1,j,k) &
. . .
287.
       1 Gg b gb
                           ss = (s0*a(i,j,k,4)-p(i,j,k))...
288.
       1 G g b gb
                           gosa1 = gosa1 + ss*ss
289.
       1 G g b gb
                           wrk2(i,j,k) = p(i,j,k) + omega*ss
290.
       1 G g b gb--->
                           ENDDO
       1 6 b---->
291.
                          ENDDO
       1 G g---->
292
                         ENDDO
       1 G-----> !$omp end target teams distribute
293.
```

Compiler listings

- vi himeno_*_v01.lst
- Below each function are additional annotations

CUDA: k value(s) built from blockldx.x

Each thread executes complete j-loop for its i, k value(s)

CUDA: i value(s) built from threadldx.x

```
269. + 1 G-----< !$omp target teams distribute reduc...
   270.
           1 G g-----
                            DO k = 2, kmax-1
   271. + 1 G g b ----- D0 j = 2, jmax-1
   272.
           1 G g b gb---< D0 i = 2, imax-1
   273.
         1 G g b gb
                           S0 = a(i,j,k,1)*p(i+1,j,k) &
   290.
          1 G g b gb--->
                               ENDDO
   291. 1 G g b---->
                              ENDDO
   292. 1 G g----->
                             ENDDO
   293. 1 G-----> !$omp end target teams distribute
 ftn-6405 ftn: ACCEL File = himeno F v01.F90, Line = 269
   A region starting at line 269 and ending at line 293 was placed on
     the accelerator.
 ftn-6430 ftn: ACCEL File = himeno F v01.F90, Line = 270
   A loop starting at line 270 was partitioned across the thread
     blocks.
 ftn-6412 ftn: ACCEL File = himeno F v01.F90, Line = 271
   A loop starting at line 271 will be redundantly executed.
→ ftn-6430 ftn: ACCEL File = himeno F v01.F90, Line = 272
   A loop starting at line 272 was partitioned across the 128 threads
     within a threadblock.
```

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Performance with the first kernel

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- Original performance:
 - \$> make VERSION=00
 - \$> qsub submit.wlm

- Gosa : 0.137835972438443027E-02
- MFLOPS : 3750.2014219750636

• With one OpenMP device kernel:

- \$> module load craype-accel-nvidia60
- \$> module switch perftools-lite-loops perftools-lite-gpu
- \$> make VERSION=01 OMP=yes
- \$> vi submit.wlm # change to v01
- \$> sbatch submit.wlm

Gosa : 0.137835972438390725E-02

MFLOPS : 2413.6436835812342

Now we use the gpu version

- The code is slower
 - Why?
 - How do we find out?

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22

more Compiler listings



To learn more, use command: explain ftn-6418

yes, as we expected

Over-cautious: compiler worried about halos; could specify map(from:wrk2)

```
+ 1 G----- !$omp target teams distribute reduc...
 270.
         1 G g-----
                          DO k = 2, kmax-1
 271. + 1 = 6 = b - - - < D0 = 2, jmax - 1
 272. 1 G g b gb---< D0 i = 2,imax-1
 273. 1 G g b gb S0 = a(i,j,k,1)*p(i+1,j,k) &
 290.
         1 G g b gb--->
                            ENDDO
         1 G g b---->
 291.
                           ENDDO
 292.
       1 G g---->
                          ENDDO
 293.
         1 G-----> !$omp end target teams distribute
LData movements:
ftn-6418 ftn: ACCEL File = himeno F v01.F90, Line = 269
 If not already present: allocate memory and copy whole array "p" to
   accelerator, free at line 293 (acc copyin).
<identical messages for a,b,c,wrk1,bnd>
ftn-6416 ftn: ACCEL File = himeno F v01.F90, Line = 269
If not already present: allocate memory and copy whole array "wrk2"
   to accelerator, copy back at line 293 (acc copy).
```

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Step 1a: data scoping

- data map clauses
 - compiler will do automatic analysis
 - usually correct
 - but can be over-cautious

• advice:

- only use clauses if compiler over-cautious
- explicit data clauses will interfere with data directives at next step

```
gosa1 = 0d0
!$omp target teams distribute &
!$omp& reduction(+:gosa1) &
!$omp& private(i,j,k,so,ss) & #optional
!$omp& map(to:a,b,c,bnd,wrk1) &
!$omp& map(to:p) map(from:wrk2)
DO k = 2, kmax-1
DO j = 2, jmax-1
 DO i = 2,imax-1
  s0 = a(i,j,k,1) * p(i+1,j,k) &
    <etc...>
     ss = (s0*a(i,j,k,4) - p(i,j,k)) * &
          bnd(i,j,k)
  gosa1 = gosa1 + ss*ss
  wrk2(i,j,k) = p(i,j,k) + omega*ss
  ENDDO
 ENDDO
ENDDO
!$omp end target teams distribute
```

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Performance with the first kernel

Original performance:

: 0.137835972438443027E-02 Gosa

MFLOPS 3750.2014219750636

With one OpenMP device kernel:

- \$> make VERSION=01a OMP=yes
- \$> vi submit.wlm # change to v01a
- \$> sbatch submit.wlm

: 0.137835972438**390725**E-02 Gosa

MFLOPS 2568.1106308604203

- Why?
- How do we find out?

The code is still slower

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Searching for performance issues



- First we look at the compiler feedback
 - Are the loops scheduling sensibly?
 - We could have a look again, but the answer is "yes".
- Next, we look at the runtime commentary
 - An event-by-event record of application execution
 - Tools from Cray and NVIDIA

- Cray runtime commentary:
 - Compile with CCE (no special options)
 - Set environment variable: CRAY_ACC_DEBUG=2
 - **Re-run** the executable
 - The commentary is written to STDERR

Cray runtime commentary

- CRAY_ACC_DEBUG=2
- Copy in/out is observed for every iteration step

Allocation and copy data to the accelerator

Copy back and release data from the accelerator

large arrays

- How often we do that?
 - Could we reduce it?

```
ACC: Start transfer 16 items from himeno F v01a.F90:153
ACC:
           allocate, copy to acc 'a' (136855584 bytes)
           allocate, copy to acc 'b' (102641688 bytes)
ACC:
           allocate, copy to acc 'bnd' (34213896 bytes)
ACC:
ACC:
           allocate, copy to acc 'c' (102641688 bytes)
           allocate, copy to acc 'gosa1' (8 bytes)
ACC:
           allocate, copy to acc 'imax' (4 bytes)
ACC:
ACC:
           allocate 'wrk2' (34213896 bytes)
ACZ:
           allocate reusable, copy to acc <internal> (4 bytes)
           allocate reusable <internal> (1008 bytes)
ACC: End transfer (to acc 444780680 bytes, to host 0 bytes)
ACC: Execute kernel jacobi  L153 1 blocks:126 threads:
             128 asynctauto) from himeno F v01a.F90:153
ACC: Wait async(auto) from himeno F v01a.F90:153
ACC: Start transfer 16 items from himeno F v01a.F90:153
          free 'a' (136855584 bytes)
ACC:
ACC:
           free 'b' (102641688 bytes)
ACC:
           free 'bnd' (34213896 bytes)
ACC:
           free 'c' (102641688 bytes)
ACC:
           copy to host, free 'gosa1' (8 bytes)
ACC:
           free 'imax' (4 bytes)
           copy to host, free 'wrk2' (34213896 bytes)
ACC:
ACC:
           done reusable <internal> (4 bytes)
           done reusable <internal> (@ bytes)
ACC:
ACC: End transfer (to acc 0 bytes, to host 34213904 bytes)
```

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Searching for performance issues (II)



NVIDIA Compute Profiler:

- Event-by-event timing information
- Compile with CCE (no special options)
- Set environment variable: COMPUTE_PROFILE=1
- Run the executable
- The commentary is written to file cuda_profile_0.log

every iteration step

Advice:

Don't set both:
 CRAY_ACC_DEBUG with
 COMPUTE PROFILE

large arrays, need long to copy

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Step 2: Optimising data movements



- Within jacobi routine
 - data-sloshing: all arrays are copied to GPU and back at every loop iteration
- Need to establish data region outside the iteration loop
 - Then data can remain resident on GPU for entire call
 - reused for each iteration without copying to/from host
 - Must accelerate all loopnests processing the arrays
 - Even if it takes negligible compute time
 - Must still accelerate for data locality
 - This can be a lot of work
 - Performance of the kernels is irrelevant.
 - A major productivity win for OpenMP compared to low-level languages
 - You can accelerate a loopnest with one directive; usually no need for tuning
 - You don't have to handcode a new CUDA kernel

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Step 2: Structure of the jacobi routine

- data region spans iteration loop
 - includes both CPU and accelerator code
 - need explicit data clauses
 - no automatic scoping
 - requires knowledge of app
 - wrk2 now a scratch array
 - does not need copying

map(alloc:wrk2) instead of map(from:wrk2)

Additional kernel to proceed everything on accelerator

```
SUBROUTINE jacobi(nn,gosa)
!$omp target data map(tofrom:p) &
       map(to:a,b,c,wrk1,bnd) &
!$omp&
       map(alloc:wrk2)
!$omp&
   iter ip: DO loop = 1,nn
      gosa = 0d0
!$omp target teams distribute <clauses>
      ! compute stencil: wrk2, gosa from p
      <stencil loopnest>
!$omp end target teams distribute
!$omp target teams distribute
     ! copy back wrk2 into p
      <copy loopnest>
!$omp end target teams distribute
   ENDDO iter lp
!$omp end target data
END SUBROUTINE jacobi
```

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Step 2: Performance with the first kernel



Original performance:

Gosa : 0.137835972438443027E-02

MFLOPS : 3750.2014219750636

With one OpenMP data region:

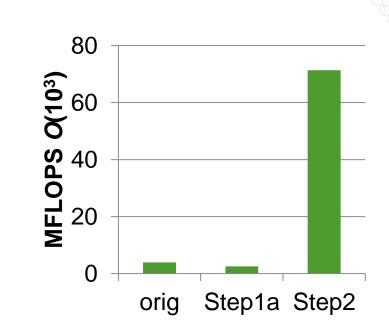
\$> make VERSION=02 OMP=yes

\$> vi submit.wlm # change to v02

\$> sbatch submit.wlm

Gosa : 0.137835972438390725E-02

MFLOPS : 71339.340223725318



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CRAY_ACC_DEBUG=2

```
CRAY
```

```
<Start of data region>
ACC: Start transfer 7 items from himeno F v02.F90:276
           allocate, copy to acc 'a' (136855584 bytes)
ACC:
           allocate, copy to acc 'b' (102641688 bytes)
ACC:
           allocate, copy to acc 'bnd' (34213896 bytes)
ACC:
ACC:
           allocate, copy to acc 'c' (102641688 bytes)
ACC:
           allocate, copy to acc 'p' (34212896 bytes)
           allocate, copy to acc 'wrk1' (34213896 bytes)
ACC:
ACC:
           allocate 'wrk2' (34213896 bytes)
ACC: End transfer (to acc 444780648 bytes, to host 0 bytes)
<For each loop iteration... (see next slide)>
<After iteration loop finishes, close of data region>
ACC: Wait async(auto) from himeno F v02.F90:329
ACC: Start transfer 7 items from himeno F v02.F90:329
          free 'a' (136855584 bytes)
ACC:
ACC:
          free 'b' (102641688 bytes)
          free 'bnd' (34213896 bytes)
ACC:
ACC:
          free 'c' (102641688 bytes)
ACC:
          copy to host, free 'p' (34213896 bytes)
ACC:
          free 'wrk1' (34213896 bytes)
          free 'wrk2' (34213896 bytes)
ACC:
ACC: End transfer (to acc 0 bytes, to host 34213896 byt)
```

Huge data movement at beginning and end of data region

```
<each iteration loop contains e.g.>
ACC: Wait async(auto) from himeno F v02.F90:311
ACC: Start transfer 7 items from himeno F_v02.F90:311
          copy to host, free 'gosa1' (8 bytes)
ACC:
          free 'imax' (4 bytes)
ACC:
ACC:
          free 'imax' (4 bytes)
ACC:
          free 'kmax' (4 bytes)
ACC:
          free 'omega' (8 bytes)
          done reusable <internal> (4 bytes)
ACC:
ACC:
          done reusable <internal> (0 bytes)
ACC: End transfer (to acc 0 bytes, to host 8 bytes)
ACC: Start transfer 3 items from himeno F v02.F90:313
ACC:
          allocate, copy to acc 'imax' (4 bytes)
          allocate, copy to acc 'imax' (4 bytes)
ACC:
ACC:
          allocate, copy to acc 'kmax' (4 bytes)
ACC: End transfer (to acc 12 bytes, to host 0 bytes)
```

Almost no data movement during iterations

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In summary

- We ported the parts of the Himeno code to the GPU
 - Add data region to avoid data transfers
 - 2 OpenACC kernels (only 1 significant for compute performance)
 - 1 data region
 - 3 directive pairs for 70 lines of Fortran/C
 - Correctness was also frequently checked
- First step was localizing suitable kernels
- First optimization of data movement
- Next steps
 - Profiling application and accelerator data
 - Checking kernels are scheduling sensibly
 - Look at kernel optimisation

