



# Is OpenMP 4.5 Target Off-load Ready for Real Life? A Case Study of Three Benchmark Kernels

Jose M. Monsalve Diaz (UDEL), Gabriele Jost  
(NASA), Sunita Chandrasekaran (UDEL) and  
Sergio Pino(ex-UDEL)

September 25, 2018  
Talk at OpenMPCon, 2018

# Outline

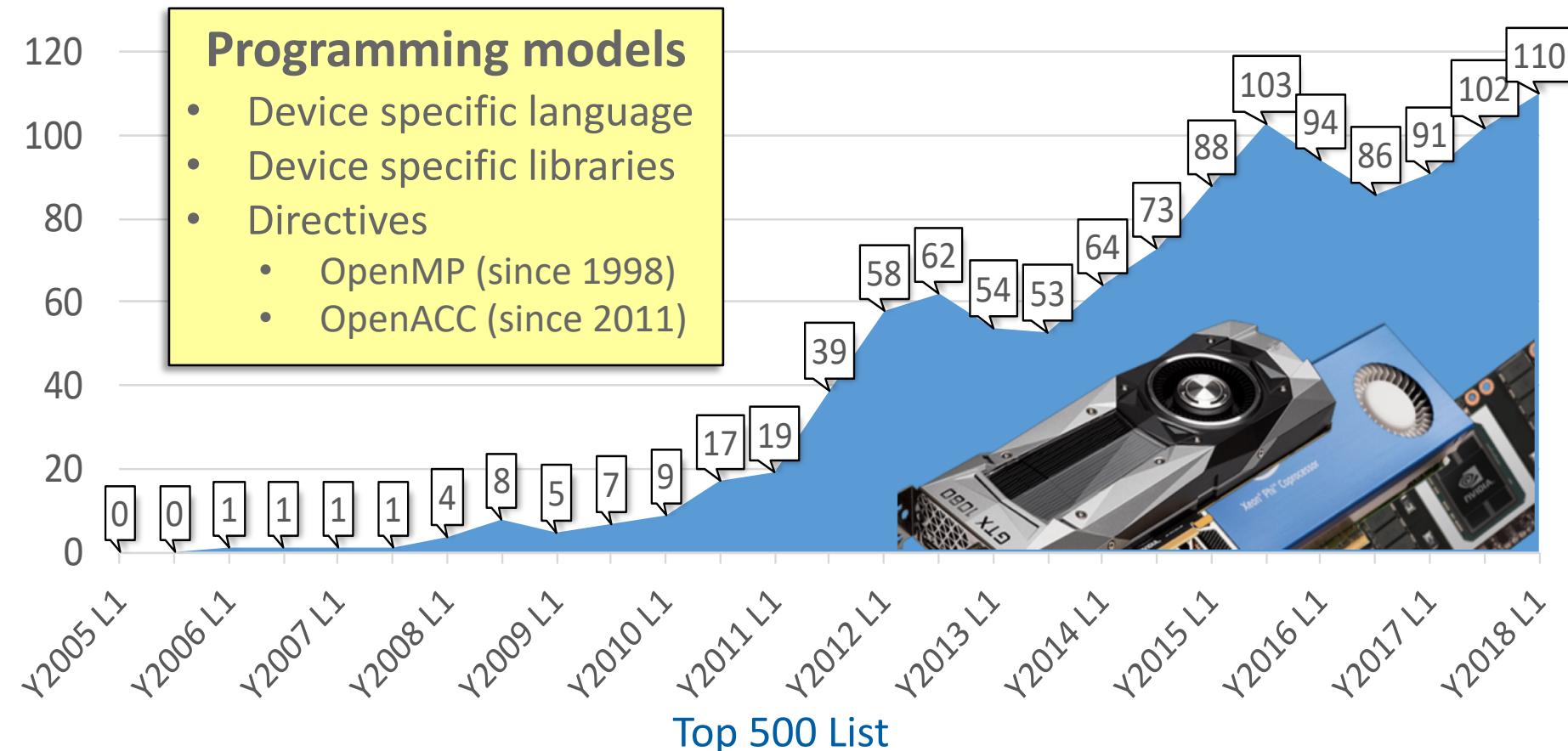


- Introduction
  - The OpenMP Target Concept
- Benchmark implementations
  - Benchmark descriptions
  - Porting OpenACC to OpenMP 4.5
- Performance Analysis:
  - Comparing compilers
  - Comparing hardware
  - Comparing OpenMP 4.5 to OpenACC
- Summary and Conclusions
- Discussion

# Introduction



## Number of systems with Accelerator devices in the Top500 list





# OpenMP and device offloading

## #pragma omp...

### OpenMP 4.0

target [data]  
declare target  
target update  
    simd  
declare simd  
    loop simd  
parallel loop simd  
    teams

distribute [simd]  
distribute parallel for [simd]  
teams distribute [simd]  
teams distribute parallel for [simd]  
target teams  
target teams distribute [simd]  
target teams distribute parallel for [simd]  
... and other API Calls ...

### OpenMP 4.5

taskloop  
taskloop simd  
target enter data  
target exit data  
    target simd  
... Other API Calls ...

### OpenMP 5.0

allocate  
declare mapper  
Memory spaces  
parallel loop  
    teams loop  
... Other API Calls ...



# Important OpenMP Constructs and Clauses

## #pragma omp target or !\$omp target

- Create data environment and execute code region on the device

## #pragma omp target map(*map-type: list*)

- Map a variable to/from the device data environment

## #pragma teams

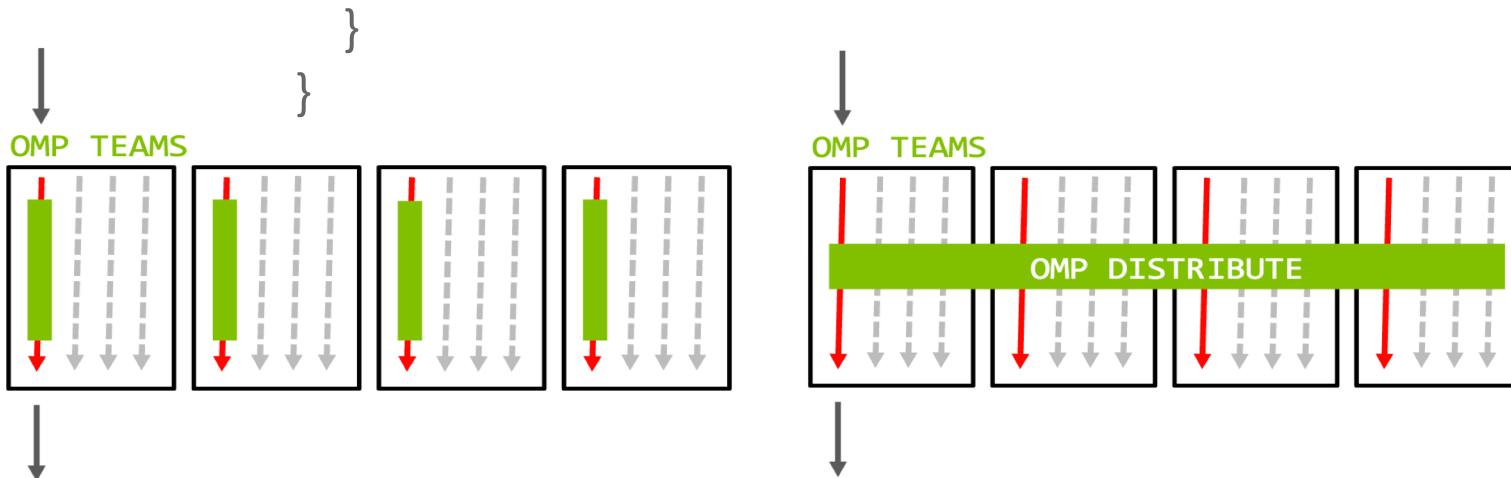
- Start kernel on the GPUs

## #pragma teams distribute, parallel for, simd

- Distribute the work across the teams and threads within each team

# Laplace Kernel Example

```
#pragma omp target teams distribute
    for( int j = 1; j < n-1; j++ ) {
        #pragma parallel for reduction(max:error)
            for( int i = 1; i < m-1; i++ ) {
                Anew[j][i] = 0.25 * ( A[j][i+1] + A[j][i-1] +
                                         A[j-1][i] + A[j+1][i] );
                error = fmax( error, fabs(Anew[j][i] - A[j][i]) );
            }
        }
    }
```



For more details check out the presentation by Jeff Larkin, Nvidia:

<http://on-demand.gputechconf.com/gtc/2016/presentation/s6510-jeff-larkin-targeting-gpus-openmp.pdf>

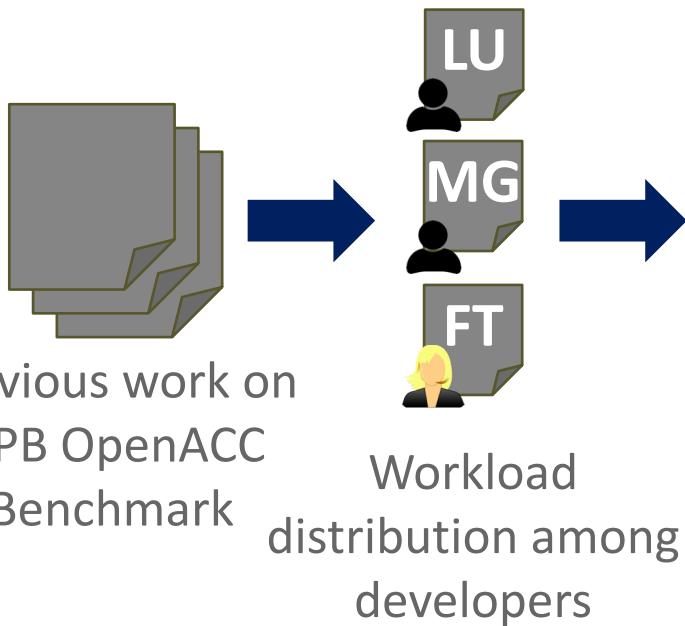


# NPB Benchmark Descriptions

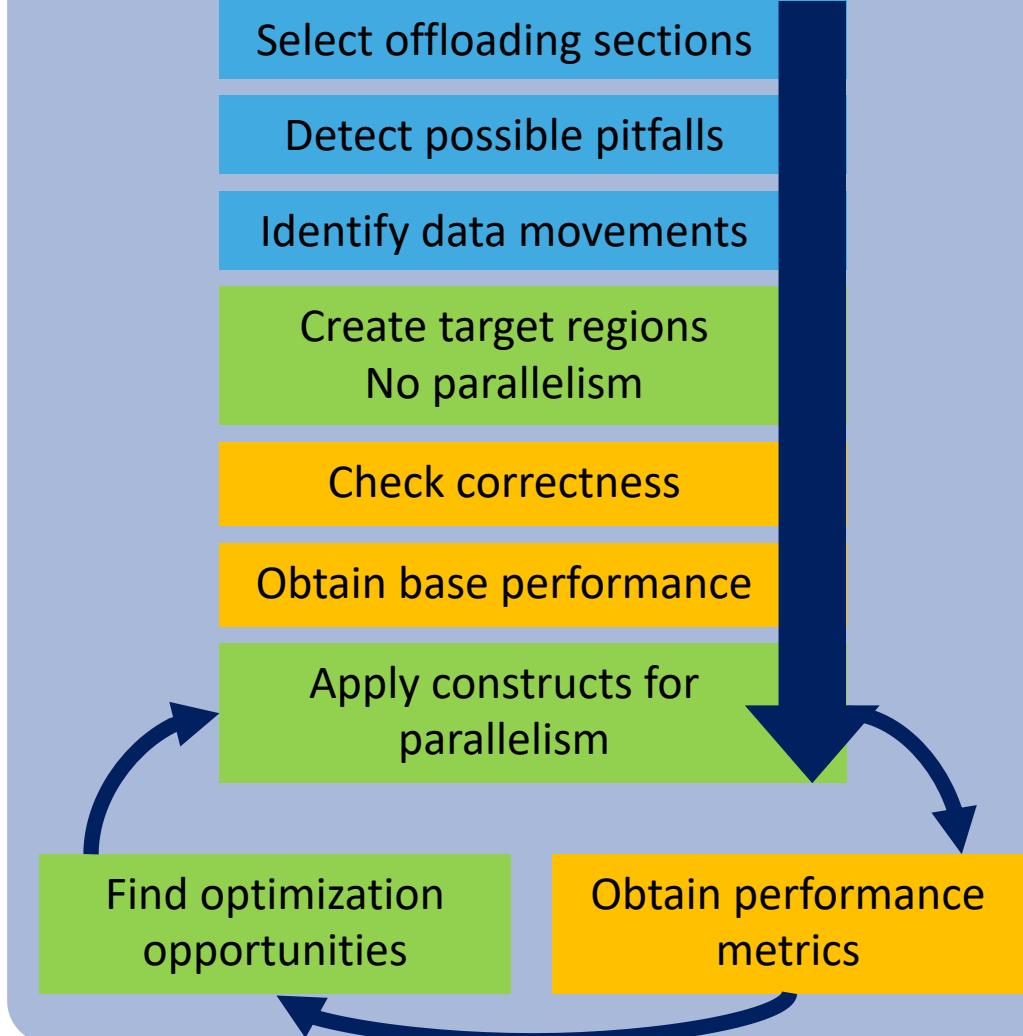
- **FT** = Discrete 3D Fast Fourier Transform
  - Requires all-to-all data transfers
  - Compiler Challenges:
    - Usage of complex data structures required manually handling real and imaginary parts separately; function calls in inner loops benefit from manual inline of function calls
- **LU-HP** = Lower-Upper Gauss Seidel Solver using a hyperplane method
  - A pipelined algorithm requires explicit thread-to-thread synchronization, which is not suitable for device execution
  - Compiler Challenges:
    - Data layout is not optimal for device execution; shared array data structures increase data transfer
- **MG** = Multi-Grid Solvers on a sequence of meshes
  - Requires long and short distance data transfers between grids
  - Memory intensive
  - Compiler Challenges:
    - 3D data structures required manual linearization
- NPB benchmark offers different classes (Problem size) – S thru E

# Development methodology

## Initial steps

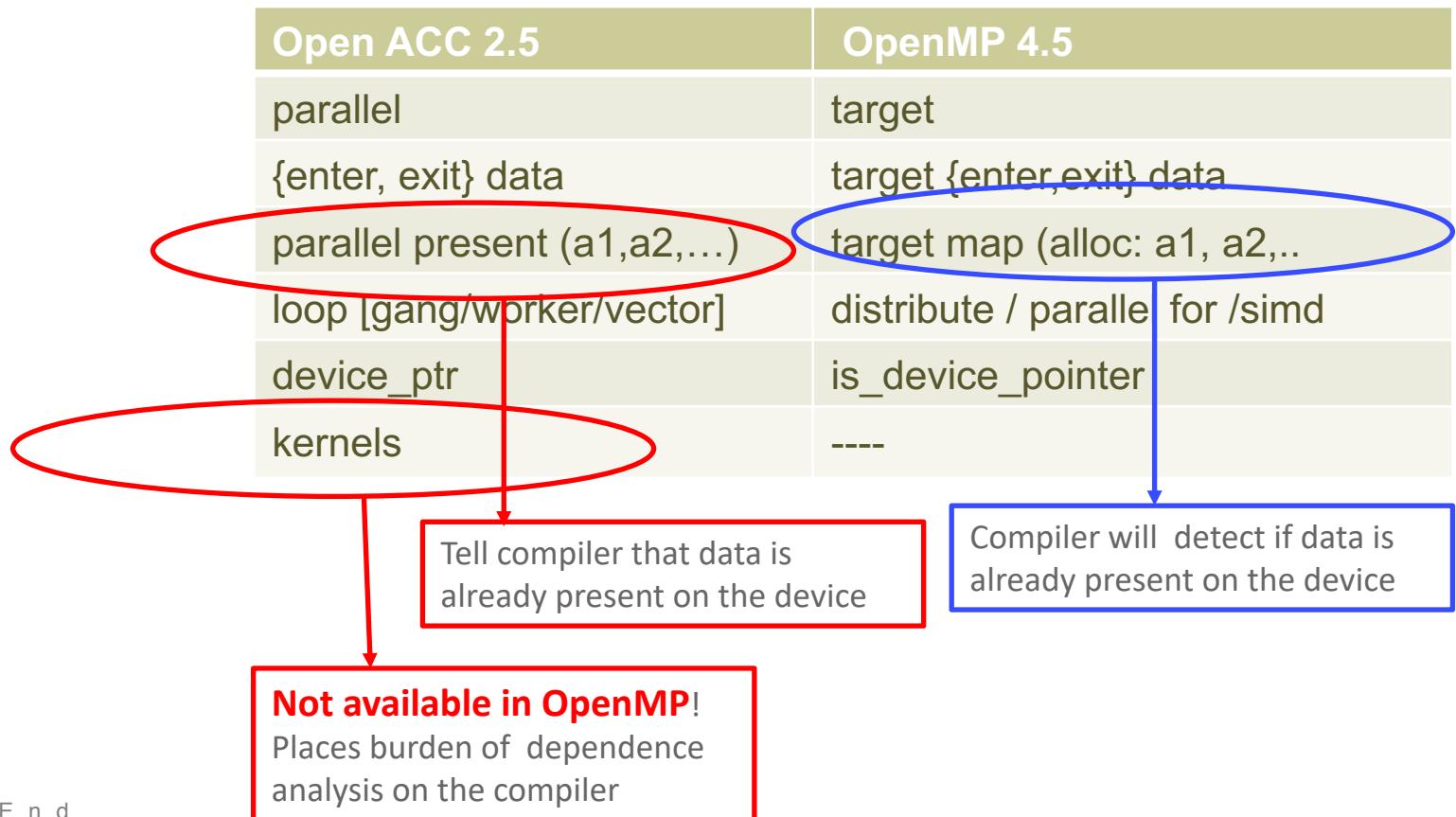


## OpenMP offloading development cycle



# General Implementation Strategy: Translating OpenACC to OpenMP

- Start out with the existing NPB 2.5 OpenACC Implementation developed in 2014 by Xu et al. (see Ref 1.)
- Translate OpenACC to OpenMP 4.5 matching constructs if available

Open ACC 2.5	OpenMP 4.5
parallel	target
{enter, exit} data	target {enter,exit} data
parallel present (a1,a2,...)	target map (alloc: a1, a2,..)
loop [gang/worker/vector]	distribute / parallel for /simd
device_ptr	is_device_pointer
kernels	----
	
<p>Tell compiler that data is already present on the device</p>	
<p>Not available in OpenMP! Places burden of dependence analysis on the compiler</p>	
<p>Compiler will detect if data is already present on the device</p>	



# FT Implementation

3D partial differential equation using an Fast Fourier Transform (FFT )

- Complex data:
  - Treat real and imaginary parts separately as in OpenACC
- Many function calls in inner loops
  - Manually inline function calls as in OpenACC

```
#pragma acc parallel num_gangs(d3) vector_length(128) \
    present(gty1_real,gty1_imag,gty2_real,gty2_imag,\ 
        u1_real,u1_imag,u_real,u_imag)
#pragma omp target map ( alloc: u1_real, u1_imag, u_real, u_imag)\ 
    map(from: gty1_real, gty1_imag, gty2_real, gty2_imag)
{
#pragma acc loop gang independent
#pragma omp teams distribute collapse(2)
    for (k = 0; k < d3; k++) {}
#pragma acc loop vector independent
    for(l = 1; l <= logd1; l += 2){
#pragma omp parallel for collapse(2) private(i11, i12, i21, i22, uu1_real, uu1_imag,
x11_real, x11_imag, x21_real, x21_imag, temp_real, temp_imag)
        for (i1 = 0; i1 <= li - 1; i1++) {
            for (k1 = 0; k1 <= lk - 1; k1++) {
                ...
                gty2_real[k][i21+k1][j] = x11_real + x21_real;
                ...
                temp_real = x11_real - x21_real;
                gty2_real[k][i22+k1][j] = (uu1_real)*(temp_real) - 
                    (uu1_imag)*(temp_imag);
            }
        }
    }
}
```



# LU-HP Implementation

Lower-Upper Gauss Seidel Solver using a hyperplane method

- Compiler Challenges:

- Array privatization
- Change data layout to enable memory coalescing
- Manual loop unrolling

```
#pragma acc parallel num_gangs(d3) vector_length(128) \
    present(gty1_real,gty1_imag,gty2_real,gty2_imag,\
        u1_real,u1_imag,u_real,u_imag)
#pragma omp target teams map (alloc: a, b, c, d, u, indxp, jndxp, rho_i, qs) \
    num_teams((npl+127)/128)
{
#pragma omp distribute parallel for private( tmp1, tmp2, i, j, k)
for (n = 1; n <= npl; n++) {
    j = jndxp[1][n];
    i = indxp[1][n];
    k = 1 - i - j;
    tmp1 = rho_i[k][j][i];
    tmp2 = tmp1 * tmp1;
    d[0][0][n] = 1.0 + dt * 2.0 * ( tx1 * dx1 + ty1 * dy1 + tz1 * dz1 );
    d[0][1][n] = -dt * 2.0
        * ( tx1 * r43 + ty1 + tz1 ) * c34 * tmp2 * u[1][k][j][i];
    d[1][1][n] = 1.0
        + dt * 2.0 * c34 * tmp1 * ( tx1 * r43 + ty1 + tz1 )
        + dt * 2.0 * ( tx1 * dx2 + ty1 * dy2 + tz1 * dz2 );
```



# MG Implementation

Multi-Grid Solvers on a sequence of meshes

- Long and short distance data transfers between grids; memory bandwidth intensive
- Compiler Challenges:
  - 3D data structures required manual linearization

```
#define I3D(array,n1,n2,i3,i2,i1) (array[(i3)*n2*n1 + (i2)*n1 + (i1)])  
  
r1 = (double*)acc_malloc(n3*n2*n1*sizeof(double))  
r1 = (double*)omp_target_alloc(n3*n2*n1*sizeof(double), omp_get_default_device());  
...  
#pragma acc data deviceptr(u1,u2), present(ou[0:n3*n2*n1]),  
    present(ov[0:n3*n2*n1], or[0:n3*n2*n1])nt n3  
#pragma acc parallel num_gangs(n3-2) num_workers(8) vector_length(128)  
#pragma omp target map(tofrom: ou[0:n3*n2*n1]) map(tofrom: ov[0:n3*n2*n1])  
map(tofrom: or[0:n3*n2*n1]) is_device_ptr(u1, u2)  
#pragma acc loop gang independent  
#pragma omp teams distribute  
for (i3 = 1; i3 < n3-1; i3++) {  
#pragma acc loop worker independent  
#pragma omp parallel for collapse(2)  
for (i2 = 1; i2 < n2-1; i2++) {  
#pragma acc loop vector independent  
for (i1 = 0; i1 < n1; i1++) {  
    I3D(u1, n1, n2, i3, i2, i1) = I3D(ou, n1, n2, i3, i2-1, i1)  
    + I3D(ou, n1, n2, i3, i2+1, i1)  
    + I3D(ou, n1, n2, i3-1, i2, i1)  
    + I3D(ou, n1, n2, i3+1, i2, i1); }}}  
NASA High End  
Computing  
Capability
```



# Evaluation Environment

	Titan	Summit	Summitdev
System	Cray XK7	IBM AC922	IBM S822LC
Nodes	6274	9216	54
CPU	16 cores AMD Opteron 6274	22 Cores IBM POWER9	20 cores IBM POWER8
Accelerators	1 NVIDIA K20X	4 NVIDIA P100	6 NVIDIA V100

## Challenge for our study:

- Different set of compilers available on different platforms
  - Each behaving differently (correctness is not always portable)
- What do we compare?
  - Selection of systems with overlapping compilers
    - Support for OpenMP Offloading in HPC system is still low despite compilers support
    - Clang trunk bug with *math.h* and host specific asm code



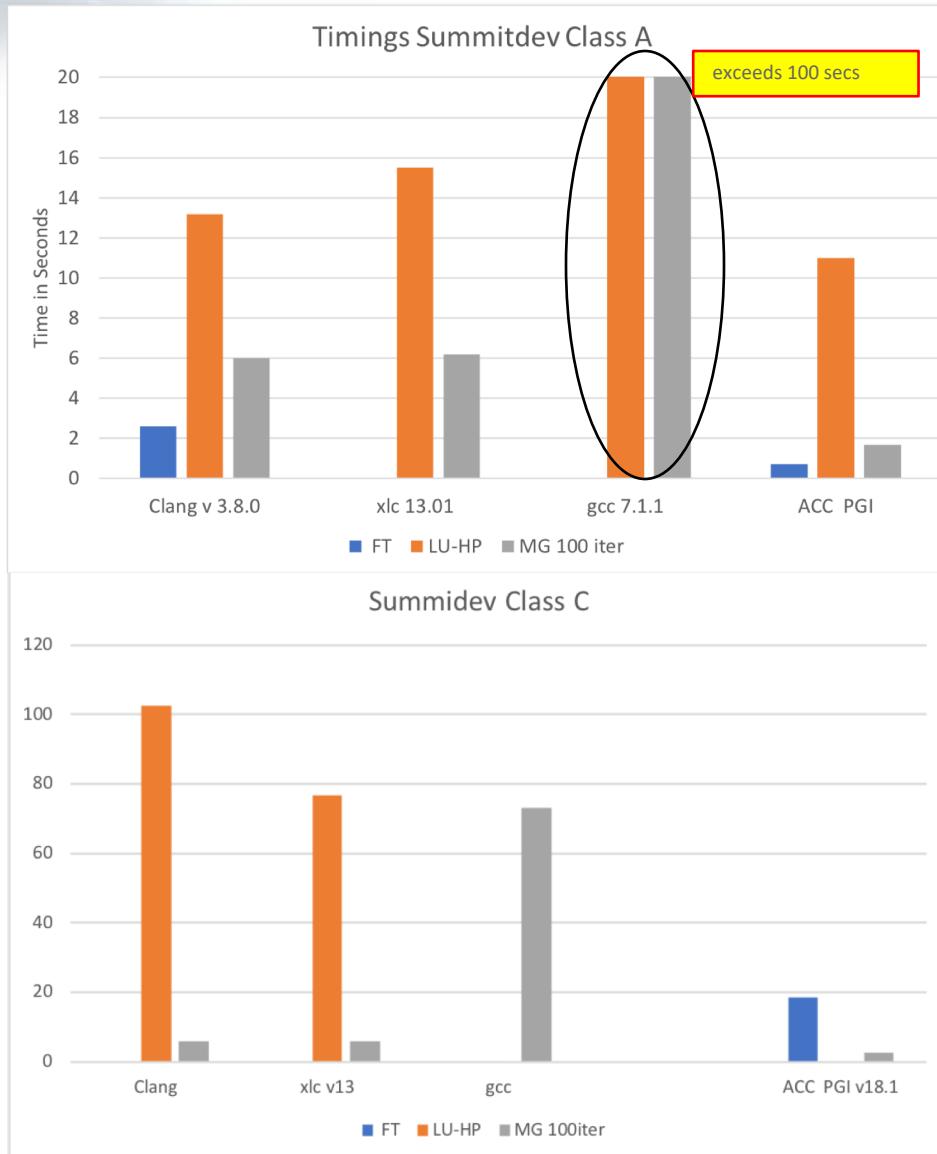
# Compilers

	Titan	Summit	Summitdev
GCC	-	-	7.1.1
PGI	18.5	18.3	18.4
CCE	8.7.3	-	-
CLANG/LLVM	-	CORAL 3.8.0	CORAL 3.8.0
XLC	-	16.1.0	13.1.0

## Challenge for our study:

- Different set of compilers available on different platforms
  - Each behaving differently (correctness is not always portable)
- What do we compare?
  - Selection of systems with overlapping compilers
    - Support for OpenMP Offloading in HPC system is still low despite compilers support
    - Clang trunk bug with *math.h* and host specific asm code

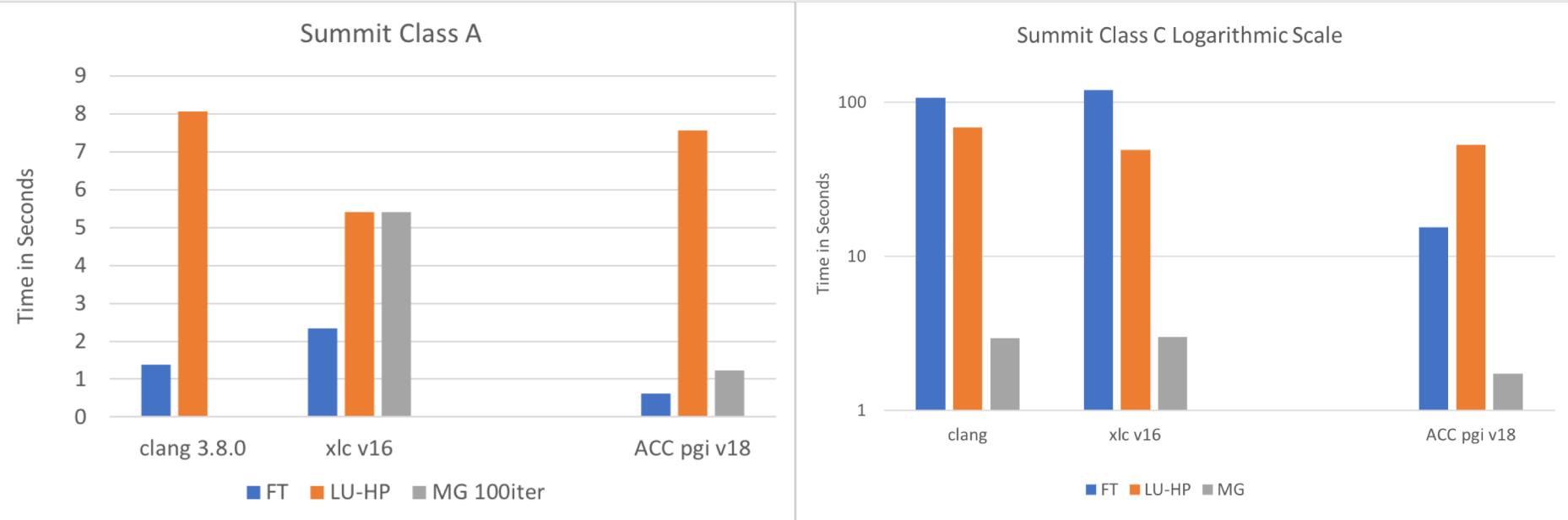
# Run time on Summitdev



- Observations:

- Runtimes of XL and Clang is quite similar
- xlc V13 failed verification for FT
- GCC 7.1.1 low performance
- PGI-OpenACC 18.1 shows relatively better performance
- PGI supports OpenMP 4.5 in their LLVM compiler, but there is no offload support yet
- Class C FT and LU-HP fail due to memory

# Run time on Summit

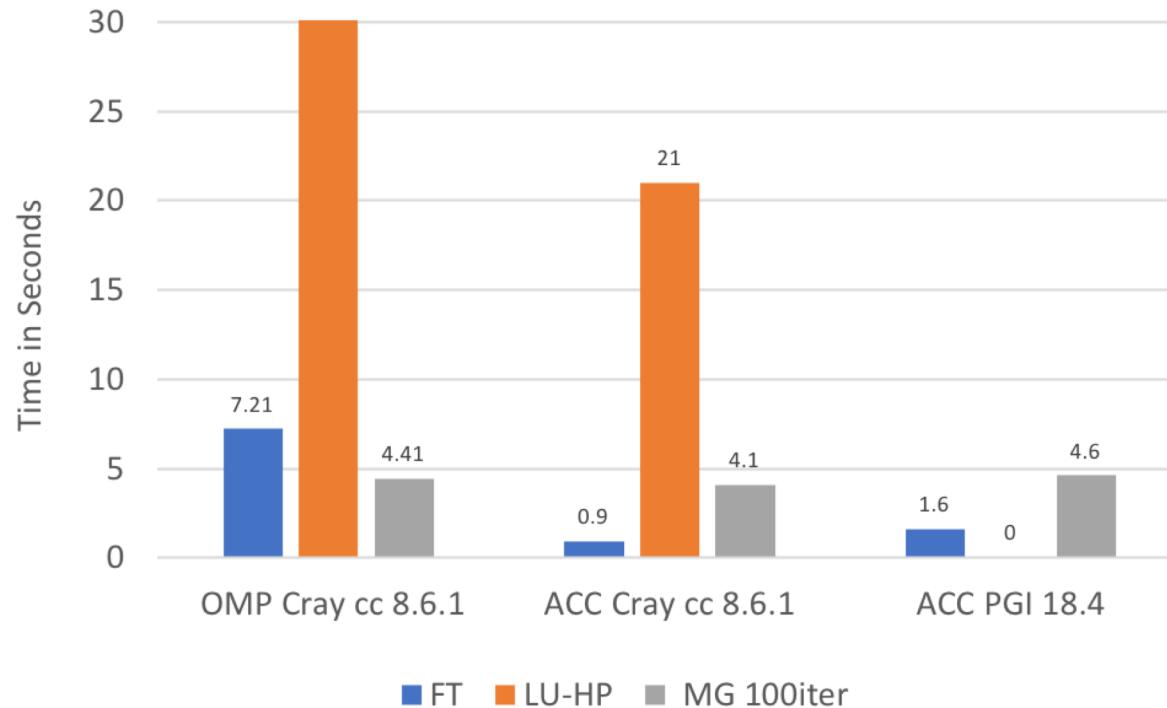


- Summit is currently unavailable due to acceptance effort ...
- PGI does not support OMP Offloading yet
- Original OpenACC employs "*pragma acc kernels*" which is not available in OpenMP 4.5

# Run time on Titan



Timings Titan Class A



- Observations:
  - For LU-HP and FT OpenACC significantly outperforms OpenMP 4.5
  - Only for MG OpenMP 4.5 can keep up with OpenACC



# Comparing Compilers: MG Class A xlc and gcc on Summitdev

## xlc v13

==67718== Profiling result:

Type	Time(%)	Time	Calls	Avg	Min	Max	Name
GPU activities:							
25.34%	292.94ms		810	361.65us	4.4800us	1.3336ms	__xl_resid_l679_Ol_4
20.51%	237.16ms		810	292.79us	5.2800us	1.1059ms	__xl_resid_l672_Ol_3
14.04%	162.31ms		808	200.88us	3.1680us	1.3819ms	__xl_psinv_l551_Ol_2
11.04%	127.61ms		808	157.93us	3.9040us	1.1059ms	__xl_psinv_l550_Ol_1
<b>0.00%</b>	<b>12.128us</b>		<b>14</b>	<b>866ns</b>	<b>704ns</b>	<b>1.6000us</b>	<b>[CUDA memcpy HtoD]</b>

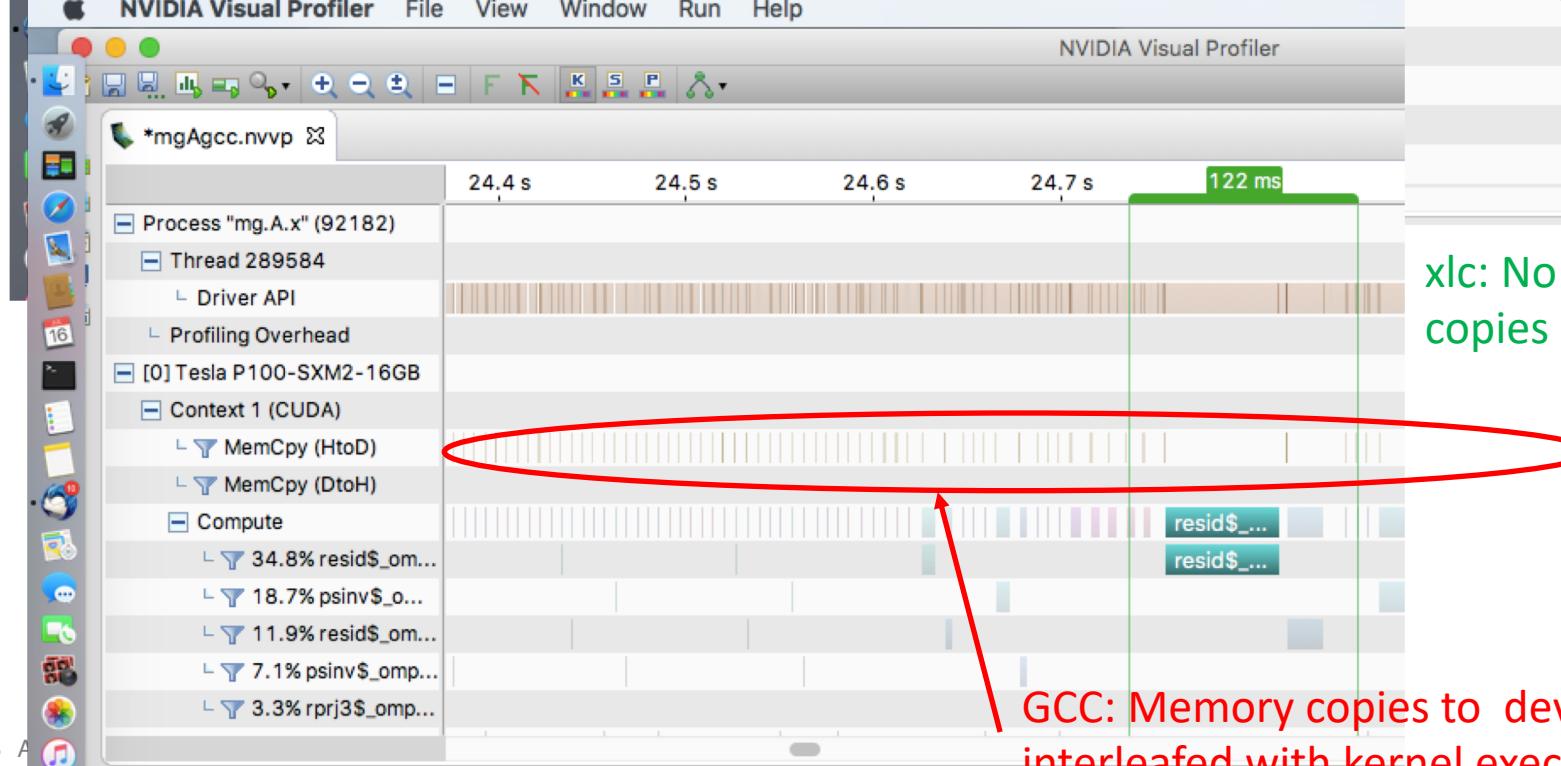
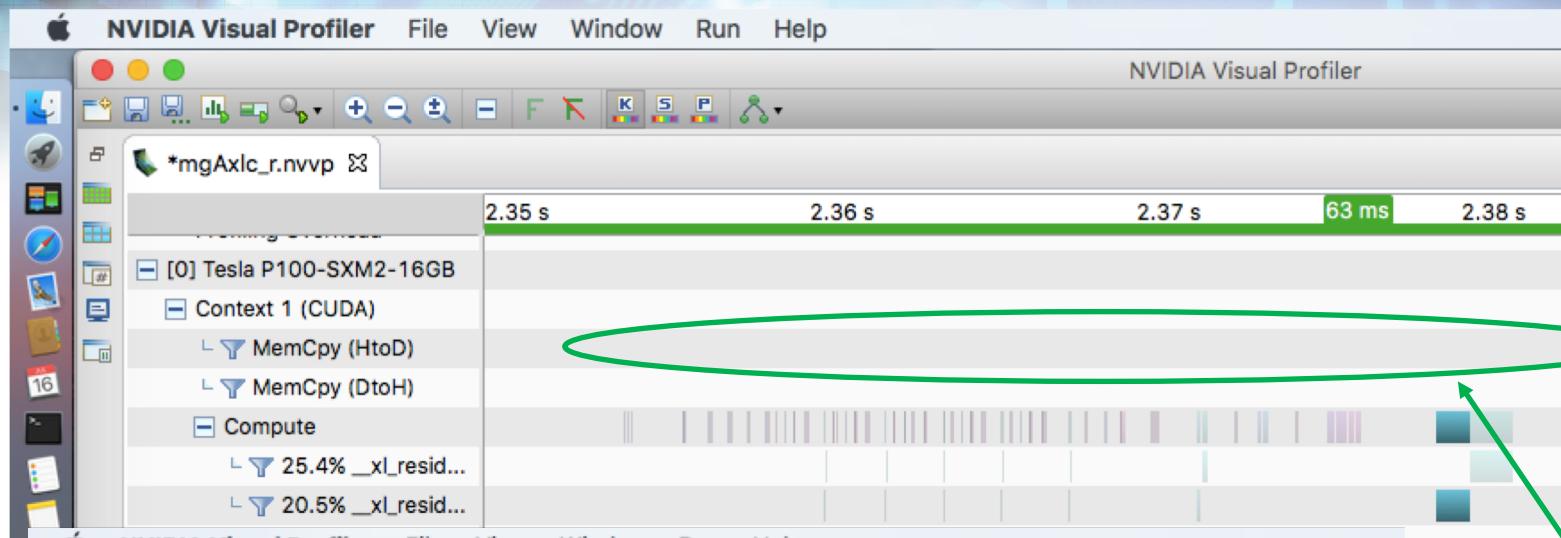
## gcc 7.1

==45872== Profiling result:

Type	Time(%)	Time	Calls	Avg	Min	Max	Name
GPU activities:							
34.59%	13.3874s		810	16.528ms	532.87us	62.394ms	resid\$_omp_fn\$44
18.63%	7.21199s		808	8.9257ms	158.34us	60.259ms	psinv\$_omp_fn\$40
11.81%	4.56960s		810	5.6415ms	384.61us	20.359ms	resid\$_omp_fn\$46
7.07%	2.73749s		808	3.3880ms	129.28us	20.859ms	psinv\$_omp_fn\$42
3.30%	1.27857s		707	1.8084ms	123.84us	8.7265ms	rprj3\$_omp_fn\$16
<b>0.33%</b>	<b>129.02ms</b>		<b>115738</b>	<b>1.1140us</b>	<b>960ns</b>	<b>2.5920us</b>	<b>[CUDA memcpy HtoD]</b>

- Observations:

- gcc shows a larger number of host-to-device data transfer
- xlc uses just a small number of asynchronous data transfers (cuMemcpyHtoDAsync)
- gcc and xlc employ different grid size, block size and number of registers per thread (see following 2 slides)



xlc: No memory copies

GCC: Memory copies to device interleaved with kernel execution



# Comparing OpenMP 4.5 vs OpenACC Performance FT on Titan

## OpenMP 4.5 + Cray cc

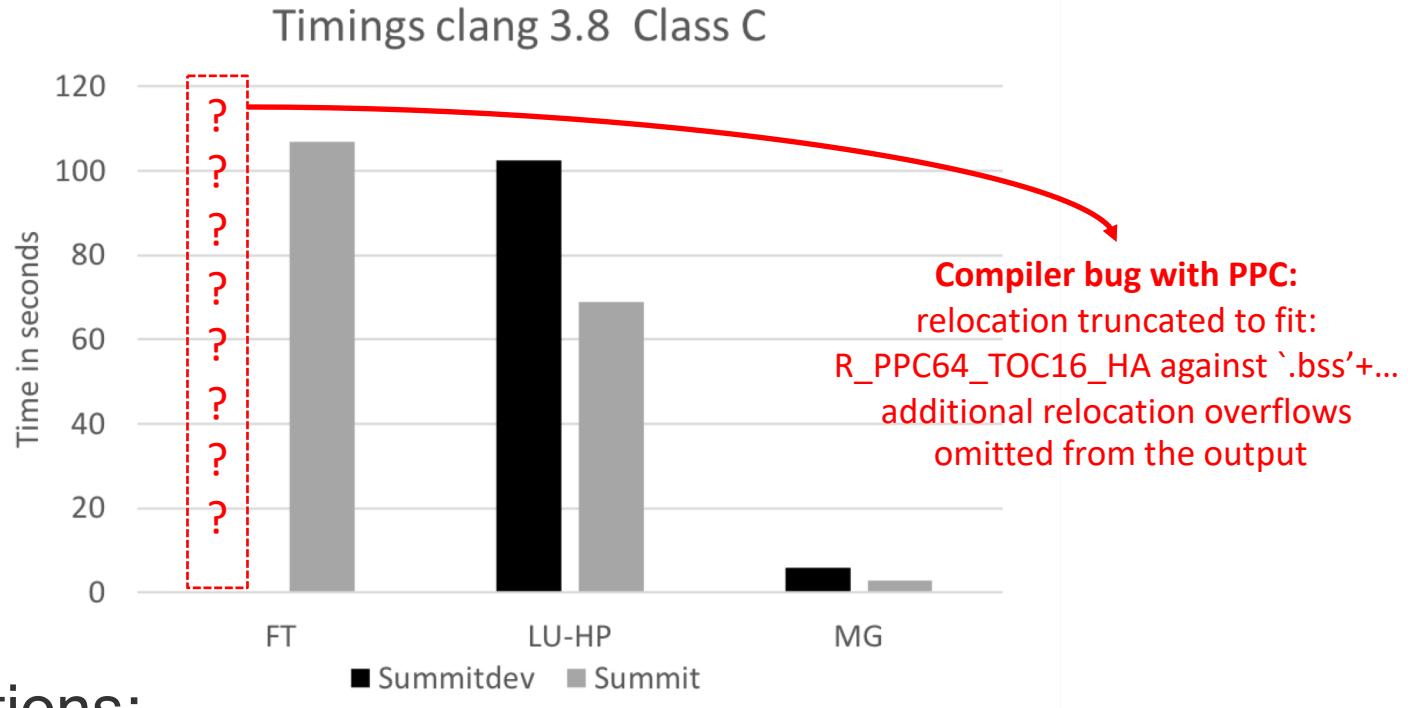
Type	Time (%)	Time	Calls	Avg	Min	Max	Name
GPU activities:	24.12%	2.20986s	6	368.31ms	368.29ms	368.36ms	cffts1_neg
	7.94%	727.67ms	58	12.546ms	1.3440us	131.51ms	[CUDA memcpy DtoH]
	3.40%	311.86ms	56	5.5689ms	928ns	41.782ms	[CUDA memcpy HtoD]
	7.91%	45.793ms	6	7.6321ms	7.5692ms	7.7115ms	cffts1_neg
	5.36%	30.988ms	6	5.1646ms	5.0361ms	5.3973ms	cffts1_neg

## OpenACC + Cray cc

Type	Time (%)	Time	Calls	Avg	Min	Max	Name
GPU activities:	32.83%	258.24ms	6	43.040ms	42.819ms	43.168ms	cffts1_neg
	13.09%	102.99ms	6	17.165ms	928ns	25.769ms	[CUDA memcpy HtoD]
	0.00%	9.9200us	6	1.6530us	1.4720us	1.9200us	[CUDA memcpy DtoH]

- Observations:
  - For each loop there are 3 kernels for OpenMP 4.5 vs 1 kernel for OpenACC
  - Data transfer to device is greatly reduced for OpenACC
  - Data transfer to host is very high in OpenMP 4.5:
    - o We had to move some arrays back to host to ensure correct execution, not necessary for OpenACC

# Comparing Hardware: Performance Summitdev vs Summit using clang 3.8



- Observations:
  - We use the same compiler version on both platforms
  - We compare impact of using 1 Nvidia P100 GPU (Summitdev) vs 1 Nvidia V100 GPU(Summit)

# Invitation to my next talk

## IWOMP Thursday 2:30pm session

OpenMP 4.5 Validation and verification Suite  
for Device offload

Jose Monsalve Diaz

Swaroop Pophale

Oscar Hernandez David E. Bernholdt Sunita Chandrasekaran



# Summary



- We described our experiences porting 3 NPB benchmarks to OpenMP 4.5 w/ offloading
- We tested our implementations on 3 different systems at OCLF
- We compared compilers, programming models and hardware
- Conclusions:
  - Evaluating 3 NPB benchmarks showed that 4.5 target offload did not lack a feature/functionality when compared with OpenACC
  - OpenMP 4.5 employs existing functionality for accelerator execution, if possible, e. g. “parallel for”, and “simd”
  - Compiler support for OpenMP would definitely benefit from further improvement
- User community:
  - Would you find it useful to have a public domain, full NPB OpenMP target implementation available?

# References



1. Xu, Rengan & Tian, Xiaonan & Chandrasekaran, Sunita & Yan, Yonghong & Chapman, Barbara. (2014). "OpenACC Parallelization and Optimization of NAS Parallel Benchmarks". 10.13140/RG.2.2.23914.41921
2. <https://www.openmp.org/specifications/> OpenMP 4.5 Specification
3. R.v.d. Pas et. al., Using OpenMP – The next Step”, MIT Press, Oct. 2017, ISBN: 9780262534789
4. S. Chandrasekaran and G. Juckeland, “OpenACC for Programmers”, Addison-Wesley Professional, September 20, 2017; ISBN-13: 978-0134694283
5. <https://www.olcf.ornl.gov/> Oak Ridge Computing Leadership Facility

## Images

<https://pixabay.com/en/user-avatar-female-blond-girl-310807/>