# HPCG and HPGMG benchmark tests on Multiple Program, Multiple Data (MPMD) mode on Blue Waters - a Cray XE6/XK7 hybrid system

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Abstract— The High Performance Conjugate Gradients (HPCG) and High Performance Geometric Multi-Grid (HPGMG) benchmarks are alternatives to the traditional LINPACK benchmark (HPL) in measuring the performance of modern HPC platforms. We performed HPCG and HPGMG benchmark tests on a Cray XE6/XK7 hybrid supercomputer, Blue Waters at National Center for Supercomputing Applications (NCSA). The benchmarks were tested on CPUbased and GPU-enabled nodes separately, and then we analyzed characteristic parameters that affect their performance. Based on our analyses, we performed HPCG and HPGMG runs in Multiple Program, Multiple Data (MPMD) mode in Cray Linux Environment in order to measure their hybrid performance on both CPU-based and GPU-enabled nodes. We observed and analyzed several performance issues during those tests. Based on lessons learned from this study, we provide recommendations about how to optimize science applications on modern hybrid HPC platforms.

# Keywords-HPC benchmark; MPMD mode; heterogeneous computing; GPU; hybrid HPC platforms

# I. INTRODUCTION

Since 1979, LINPACK benchmark (HPL) [1] has been used to measure the performance of HPC systems and the TOP500 list has been accordingly updated to rank the world's most powerful supercomputer. However, there have been many opinions criticizing whether HPL is a good metric to measure modern HPC performance or not. One of the main criticisms is its low memory to flop (Byte/Flop) ratio. Byte/Flop ratios of many applications in molecular dynamics, weather forecasting, astrophysics, particle physics, structural analysis and fluid dynamics are between  $10^{-1}$  to 10, while the ratio of HPL benchmark is less than  $10^{-3}$ . Since 2014, HPC community has shared the High Performance Conjugate Gradients (HPCG) [2,3] and High Performance Geometric Multi-Grid (HPGMG) [4,5] benchmark results in order to represent modern HPC performance for many science and engineering HPC applications, at least, in terms of Byte/Flop ratios.

This study is an extension of our continuing effort [6] to find out appropriate benchmarks for modern HPCs. In this study, we employed HPCG and HPGMG benchmarks to measure the performance of the Blue Waters system located at National Center for Supercomputing Application (NCSA). Blue Waters [7,8] is a Cray XE6/XK7 hybrid supercomputer with 22,640 CPU-based XE6 nodes and 4,228 GPU-enabled XK7 nodes. The XE6 dual-socket nodes are populated with 2 AMD Interlargos model 6276 CPU processors with a nominal clock speed of at least 2.3 GHz and 64 GB of physical memory, while the XK7 accelerator nodes are equipped with one Interlagos model 6276 CPU process and one NVIDIA GK110 "Kepler" accelerator K20X with 32 GB of CPU memory and 6 GB of GPU memory. We first tested HPCG and HPGMG on dual-socket CPU-based nodes and single-socket GPU-enabled nodes separately. After analyzing numerous configurations for the optimal performance, we moved on Multiple Program, Multiple Data (MPMD) runs to evaluate the performance on CPU-based XE6 nodes and GPU-enabled XK7 nodes together. We provide performance analyses results based on MPMD runs and our recommendations for other HPC applications on modern hybrid HPC platforms.

#### II. HPCG BENCHMARK ON BLUE WATERS

# A. General Description

The HPCG [2,3] implementation contains crucial computational and communicational patterns present in various methods for discretization and numerical solutions of PDEs, such as dense and sparse computations, dense and sparse collectives, multi-scale execution of kernels through truncated multi-grid V cycles, and data-driven parallelism for unstructured sparse triangular solves. It synthetically discretizes an elliptic Partial Differential Equation (PDE) in 3 dimensions with zero Dirichlet boundary conditions (BCs) and a synthetic right hand side (RHS) vector. A local domain,  $n_x \times n_v \times n_z$  is distributed into all available MPI ranks with a process layout  $np_x \times np_y \times np_z$ ; as a result, the global domain size becomes  $(np_x \times n_x) \times (np_y \times n_y) \times (np_z \times n_z)$ . The global sparse matrix is a symmetric positive definite matrix assembled by element-wise 27-point stencil operator. It is solved via the CG iterations including a local and symmetric Gauss-Seidel pre-conditioner, which computes a forward and a backward solve with a triangular matrix. The HPCG benchmark subsequently rewards investments in highperformance collective operations, local memory system performance and low latency cooperative threading [3]. At SC16, the 6<sup>th</sup> HPCG performance list was announced. Five systems in the top 10 are GPU- or XEON Phi-based systems, while other five systems are CPU-based systems.

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In this section, we first present HPCG benchmark results on CPU-based XE nodes or GPU-enabled XK nodes. We tested them for several different configurations in terms of local block sizes, process layouts, number of threads, and number of nodes. Based on analyses of the numerous test results, we performed MPMD runs using XE and XK nodes together for HPCG benchmark. At the end of this section, we discuss load-balancing issues between XE and XK nodes, and provide recommendations for HPC applications on hybrid systems.

# *B. HPCG on CPU-based XE nodes*

We downloaded source files of HPCG revision 3.0 from the official HPCG GitHub repository (https://github.com /hpcg-benchmark/hpcg). We first built an executable with gcc/4.9.3 and cray-mpich/7.3.0 on Blue Waters (i.e., HPCG-CPU-GitHub-GNU), and then tested it for a variety of MPIrank and thread combinations, and for various sizes of local blocks. Based on these results, we performed a scaling test with up to 4 % of Blue Waters XE nodes with HPCG-CPU-GitHub-GNU. In addition, we built the second executable binary with intel/16.0.3.210 with cray-mpich/7.3.0 on Blue Waters (i.e., HPCG-CPU-GitHub-Intel), and then conducted an additional scaling test with the new binary.

A single XE node on Blue Waters has 16 floating-point units (FPUs) with two integer cores sharing a single FPU (i.e., total 32 integer cores in an XE node). Depending on usage rates of FPUs, some applications may optimally perform by assigning a single core for a thread, while others may perform better with a single FPU per thread. Figure 1 and Table I show HPCG-CPU-GitHub-GNU performance for multiple threads on 4 XE nodes in weak scaling for MPI rank. Each MPI rank has 104<sup>3</sup> local blocks; therefore, the global domain size as well as memory usage decreases as the number of threads increases. The red line shows HPCG performance when we assigned one integer core per thread. The blue line shows results when one FPU is assigned for a thread. The case with 128 MPI ranks and 1 integer core per thread shows the best performance, though other cases with 64 MPI ranks and 1 integer core or FPU per thread show similar results to the best case. It means HPCG on Blue Waters shows an optimal performance when its OpenMP threads share a single FPU, not multiple FPUs. In other cases where each MPI rank uses multiple FPUs through threading, HPCG performance noticeably drops even on the same number of XE nodes.

Figure 2 and Table II show HPCG-CPU-GitHub-GNU performance for multiple threads on 4 XE nodes in strong scaling for MPI rank. We configured that the global domain has  $832 \times 416^2$  blocks for all cases. As number of threads increases, we assigned larger blocks to each MPI rank. All cases, consequently, used around 103 GB in memory that is approximately 40% of memory on 4 XE nodes. When each MPI used a single FPU, HPCG showed the best performance as observed in the previous tests. The performance of strong scaling tests in Figure 2 shows a little bit more drops with many threads than weak scaling tests in Figure 1, but the difference between weak and strong scaling tests was not significant.



Figure 1. HPCG-CPU-GitHub-GNU for multiple threads on 4 XE nodes – weak scaling for MPI rank (the same number of equations per MPI rank)

 TABLE I.
 HPCG-CPU-GITHUB-GNU RESULTS FOR MULTIPLE

 THREADS ON 4 XE NODES IN WEAK SCALING FOR MPI RANK

Core type per thread	MPI rank	Number of threads	HPCG (GFLOP/s)	Number of Equations in total	Memo ry (GB)
	128	1	21.4798	143,982,592	102.99
1 :	64	2	21.1808	71,991,296	51.50
1 integer	32	4	15.6547	35,995,648	25.75
thread	16	8	9.47234	17,997,824	12.87
uncau	8	16	4.53633	8,998,912	6.44
	4	32	2.2442	4,499,456	3.22
	64	1	20.7478	71,991,296	51.50
1 FPU	32	2	15.5168	35,995,648	25.75
per	16	4	9.50418	17,997,824	12.87
thread	8	8	4.45687	8,998,912	6.44
	4	16	2.24646	4,499,456	3.22



Figure 2. HPCG-CPU-GitHub-GNU for multiple threads on 4 XE nodes – strong scaling for MPI rank (the same number of equations in total)

Core type per thread	MPI rank	Number of threads	HPCG (GFLOP/s)	Number of Equations in total	Memo ry (GB)
	128	1	21.8334	143,982,592	102.99
1 :	64	2	21.1425	143,982,592	102.98
1 integer	32	4	15.3969	143,982,592	102.96
throad	16	8	9.95385	143,982,592	102.95
uneau	8	16	4.34153	143,982,592	102.94
	4	32	2.09014	143,982,592	102.92
	64	1	20.74	143,982,592	102.98
1 FPU	32	2	15.5413	143,982,592	102.96
per	16	4	9.45932	143,982,592	102.95
thread	8	8	4.23567	143,982,592	102.94
	4	16	2.00099	143,982,592	102.92

TABLE II. HPCG-CPU-GITHUB-GNU RESULTS FOR MULTIPLE THREADS ON 4 XE NODES IN STRONG SCALING FOR MPI RANK



Figure 3. HPCG-CPU-GitHub-GNU performance for various local block sizes on 4 XE nodes

 
 TABLE III.
 HPCG-CPU-GITHUB-GNU PERFORMANCE FOR VARIOUS LOCAL BLOCK SIZES ON 4 XE NODES

n <sub>x</sub>	$n_y$	nz	HPCG (GFLOP/s)	Number of equations in total	Memory (GB)
16	16	16	26.5972	524,288	0.38
24	24	24	24.5402	1,769,472	1.27
32	32	32	23.0211	4,194,304	3.01
40	40	40	22.9008	8,192,000	5.87
48	48	48	22.5263	14,155,776	10.14
56	56	56	22.0341	22,478,848	16.09
64	64	64	21.6028	33,554,432	24.02
72	72	72	21.7075	47,775,744	34.19
88	88	88	21.3858	87,228,416	62.41
96	96	96	21.3235	113,246,208	81.01
104	104	104	21.8639	143,982,592	102.99
112	112	112	21.8343	179,830,784	128.63
120	120	120	21.6664	221,184,000	158.20
128	128	128	21.5274	268,435,456	191.98

Figure 3 and Table III present HPCG-CPU-GitHub-GNU performance for different local block sizes on 4 XE nodes. Since HPCG in the previous tests shows the best performance with 128 MPI ranks on 4 XE nodes, we again

used the same configuration for this test. The x-axis shows number of local blocks along each direction (i.e.,  $n_x = n_y = n_z$ ) and y-axis shows HPCG results in GFLOP/s and corresponding memory usages in GB. As  $n_x$  increases, the memory usage on 4 XE nodes exponentially grows (i.e., blue line in Figure 3). However, HPCG performance (i.e., red line in Figure 3) keeps constant after sharp drops for cases with very small numbers of equations. The green dotted line in Figure 3 shows a quarter of full memory on 4 XE nodes that is the required minimum memory usage for official HPCG results. Through this test, it turns out that local block size does not significantly affect HPCG performance once it satisfies the official HPCG requirement for memory usage.

 TABLE IV.
 Scaling test for HPCG-CPU-GitHub-GNU with up to 3.6% Blue Waters XE nodes

XEs	MPI ranks	HPCG (GFLOP/s)	Number of equations in total	Memory (GB)	Effici- ency (%)
1	32	5.43104	35,995,648	25.75	100
2	64	11.355	71,991,296	51.50	105
4	128	21.9875	143,982,592	102.99	101
8	256	43.5148	287,965,184	205.99	100
16	512	86.6853	575,930,368	411.97	100
32	1024	170.412	1,151,860,736	823.94	98
50	1600	268.921	1,799,782,400	1,287.40	99
100	3200	538.826	3,599,564,800	2574.81	99
200	6400	1057.59	7,199,129,600	5149.62	97
400	12800	2154.29	14,398,259,200	10299.2	99
800	25600	4137.76	28,796,518,400	20598.50	95

 
 TABLE V.
 Scaling test for HPCG-CPU-GitHub-Intel with up to 4.6% Blue Waters XE nodes

XEs	MPI ranks	HPCG (GFLOP/s)	Number of equations in total	Memory (GB)	Effici- ency (%)
1	32	7.02902	35,995,648	25.75	100
2	64	13.0688	71,991,296	51.50	93
4	128	27.9278	143,982,592	102.99	99
8	256	55.3573	287,965,184	205.99	98
16	512	104.433	575,930,368	411.97	93
32	1024	218.144	1,151,860,736	823.94	97
64	2048	435.228	2,303,721,472	1,647.88	97
128	4096	817.343	4,607,442,944	3,295.75	91
256	8192	1646.64	9,214,885,888	6591.51	92
512	16384	3103.43	18,429,771,776	13183	86
1024	32768	6218.57	36,859,543,552	26366	86

We chose  $n_x = n_y = n_z = 104$  for a scaling test for HPCG-CPU-GitHub-GNU, and conducted it with up to 800 XE nodes (i.e., around 3.6% of Blue Waters XE nodes). Based on our previous test results, we assigned 32 MPI ranks per node (i.e., 1 MPI rank to 1 integer core); therefore, number of MPI ranks used in this test was from 32 to 25,600. The memory usage was around 40% of assigned XE nodes. As shown in Table IV, HPCG-CPU-GitHub-GNU shows an impressive scaling on Blue Waters, and it keeps more than or equal to 95% of parallel efficiency in all of the cases.

We conducted an additional scaling test for HPCG-CPU-GitHub-Intel with up to 1,024 XE nodes (i.e., around 4.6% of Blue Waters XE nodes) with the same configuration (i.e.,  $n_x = n_y = n_z = 104$  and 32 MPI/node). The range of MPI ranks is from 32 to 32,768 as shown in Table V. The parallel efficiency of HPCG-CPU-GitHub-Intel is more than or equal to 86%, which is lower than results of HPCG-CPU-GitHub-GNU. The overall performance of HPCG-CPU-GitHub-Intel, however, is much better than the performance of HPCG-CPU-GitHub-Intel, as compared in Figure 4.



Figure 4. Scaling tests for HPCG-CPU-GitHubs up to 4.6% Blue Waters XE nodes

#### C. HPCG on GPU-enabled XK nodes

The official HPCG webpage (i.e., www.hpcg-benchmark. org) provides HPCG 3.0 binary from NVIDIA optimized for NVIDIA GPUs. It requires CUDA and OpenMPI 1.6.5 libraries. However, in [6], it turned out that executable binaries built for the OpenMPI library could not perform optimally on Blue Waters. In this study, we instead employed a Cray-MPICH compatible HPCG binary provided by NIVIDA (i.e., HPCG-GPU-NVIDIA).

Each XK node on Blue Waters has one CPU processor with 32GB memory and one K20X GPU with 6 GB device memory. In this section, we assigned one MPI rank per GPU; as a result, one MPI rank was assigned to each node. Since GPUs do most of computations for HPCG, we kept an eve on GPU memory capacity for assigned XK nodes, instead of CPU memory capacity. Figure 5 and Table VI show HPCG-GPU-NVIDIA performance for various local block sizes on 8 XK nodes. The x-axis of Figure 5 is a local cubic block size along each direction (i.e.,  $n_x$ ,  $n_y$ , and  $n_z$ ). As it grows, the memory used for data exponentially increases (i.e., red line in Figure 5). The green dotted line in Figure 5 shows the minimum required memory usage for the official HPCG results that is a quarter of GPU memory of 8 XK nodes. The blue line in Figure 5 shows HPCG results. The HPCG performance linearly increases from 24<sup>3</sup> to 88<sup>3</sup> of local blocks, since communication overheads between CPU and GPU in these cases rapidly tails off. After that, HPCG results look converged into some range. One interesting observation is that memory used for cases with 208<sup>3</sup> and

 $216^3$  local cubes are larger than the total GPU memory of 8 XK nodes (i.e., 48 GB). It seems HPCG-GPU-NVIDIA somehow re-uses GPU memory during the computation. For the larger local domain than  $216^3$ , the out-of-memory error occurred. In summary, HPCG-GPU-NVIDIA shows the most optimal performance with  $128^3$ ,  $168^3$ , and  $176^3$  local blocks on 8 XK nodes.



Figure 5. HPCG-GPU-NVIDIA for various local block sizes on 8 XK nodes

TABLE VI. HPCG-GPU-NVIDIA PERFORMANCE FOR VARIOUS LOCAL BLOCK SIZES ON 8 XK NODES

			HPCG	Number of	Memory
$n_x$	$n_y$	nz	(GFLOP/s)	equations in total	(GB)
24	24	24	15.5542	110,592	0.079
32	32	32	38.661	262,144	0.188
40	40	40	55.1535	512,000	0.367
48	48	48	83.3617	884,736	0.634
56	56	56	102.637	1,404,928	1.006
64	64	64	134.636	2,097,152	1.501
72	72	72	126.136	2,985,984	2.137
80	80	80	150.928	4,096,000	2.931
88	88	88	157.044	5,451,776	3.900
96	96	96	172.29	7,077,888	5.063
104	104	104	169.535	8,998,912	6.437
112	112	112	177.864	11,239,424	8.039
120	120	120	174.563	13,824,000	9.887
128	128	128	204.912	16,777,216	11.999
136	136	136	179.439	20,123,648	14.392
144	144	144	191.406	23,887,872	17.083
152	152	152	187.779	28,094,464	20.091
160	160	160	190.252	32,768,000	23.432
168	168	168	209.338	37,933,056	27.125
176	176	176	207.152	43,614,208	31.187
184	184	184	167.078	49,836,032	35.635
192	192	192	164.078	56,623,104	40.487
200	200	200	178.81	64,000,000	45.761
208	208	208	185.213	71,991,296	51.474
216	216	216	180.76	80,621,568	57.644

With 128<sup>3</sup>, 168<sup>3</sup> and 176<sup>3</sup> local blocks, we carried out scaling tests with up to 6.1% Blue Water XK nodes (i.e., 256 XK nodes). Table VII shows HPCG results in GFLOP/s and

parallel efficiency in percentile for three cases. Figure 6 shows plots for number of XK nodes versus HPCG results for 128<sup>3</sup>, 168<sup>3</sup> and 176<sup>3</sup> local blocks. The purple dotted line describes a linear scaling. The case with 168<sup>3</sup> local blocks shows the best HPCG performance on 8, 16 and 32 XK nodes, while the case with 128<sup>3</sup> local blocks was the best on other numbers of XK nodes. The parallel efficiency is more than or equal to 83%, which is worse than the parallel efficiency of HPCG-CPU-GitHub-Intel (i.e., 92% with 256 XE nodes). We think this happens because of two reasons:

- Additional communicational overhead between the host (i.e., CPU) and the device (i.e., GPU) on XK nodes reduces the network performance for MPI communications.
- Higher HPCG performance of a single XK node (i.e., 27 GFLOP/s/node for GPU version in Table VII) than of a single XE node (i.e., 7 GFLOP/s/node for CPU version in Table V).

TABLE VII. SCALING TEST FOR HPCG-GPU-NVIDIA WITH UP TO 6.1% BLUE WATERS XK NODES

		$n_x = n_y = n_z = 128$		$n_x = n_y = n_z = 168$		$n_x = n_y = n_z = 176$	
XKs	MPI	HPCG	Effici-	HPCG	Effici-	HPCG	Effici-
21113	ranks	(GFL	ency	(GFLO	ency	(GFLO	ency
		OP/s)	(%)	P/s)	(%)	<i>P/s</i> )	(%)
1	1	27.029	100	23.544	100	23.734	100
2	2	53.353	99	46.811	99	48.960	103
4	4	103.84	96	93.085	99	95.674	101
8	8	204.87	95	209.33	111	207.19	109
16	16	397.87	92	415.67	110	412.30	109
32	32	790.06	91	827.76	110	692.05	91
64	64	1493.0	86	1347.1	89	1365.9	90
128	128	2941.0	85	2741.2	91	2724.6	90
256	256	5756.2	83	5365.2	89	5359.8	88



Figure 6. Scaling tests for HPCG-GPU-NVIDIA up to 6.1% Blue Waters XK nodes

# D. HPCG on CPU-based XE and GPU-enabled XK nodes on MPMD mode

For Multiple Program, Multiple Data (MPMD) mode, binary combinations for CPU-based XE nodes and GPUenable XK nodes should be built from the same base code. In this section, we used two new binaries provided by NVIDIA: HPCG-MPMD-CPU for XE nodes and HPCG-MPMD-GPU for XK nodes.

Since the base code is slightly different from our previous studies on XE or XK node, we briefly performed a series of single node test for different local block sizes (e.g.,  $72^3$ ,  $104^3$ ,  $128^3$  and  $128 \times 64^2$ ) and various combinations of MPI ranks and threads (e.g., 1 to 32 MPIs with a thread per an integer core on XE, and 1 to 16 MPIs on XK). Table VIII and Figure 7 shows HPCG performance of HPCG-MPMD-CPU on a single XE node and HPCG-MPMD-GPU on a single XK node. Because of different memory capacity of CPUs (i.e., 64 GB) on XE and GPU (i.e., 6 GB) on XK, some cases with a large number of MPI ranks and big local blocks did not complete due to the out-of-memory error. The best XE node performance was 8.3 GFLOP/s with  $128 \times 64^{2}$ local blocks when 16 MPI ranks with 2 threads were assigned on the XE node. The best XK node performance was 22.8 GFLOP/s with 128×64<sup>2</sup> local blocks when 14 MPI ranks are assigned to the XK node. According to comparison of the best performances, HPCG on an XK node is 2.7 times faster than HPCG on an XE node. This difference can result in a serious load imbalance between XE and XK nodes; as a result, the performance on MPMD mode can be degraded.

Our approach to minimize the load imbalance between CPU-only XE and GPU-enable XK nodes is to synchronize XE nodes' HPCG performance per MPI rank with XK nodes'. Figure 8 presents HPCG performance per MPI rank of HPCG-MPMD-CPU on an XE node and HPCG-MPMD-GPU on an XK node. As number of MPI ranks increases, HPCG performance per MPI on the XK node linearly decreases. On the XE node, HPCG performance per MPI rank linearly increases as number of threads increase up to 8 threads per MPI rank.

TABLE VIII. HPCG PERFORMANCE OF HPCG-MPMD-CPU AND HPCG-MPMD-GPU ON A SINGLE XE OR XK NODE(UNIT: GFLOP/S)

N. I	MDI	1		1		100
Node	MPI	Threads	nx=ny=	nx=ny=	nx=ny=	nz=128,
type ranks		1 m caus	nz=72	nz=104	nz=128	ny=nz=64
	1	32	1.0335	1.13907	1.39787	1.43499
	2	16	2.28724	2.2585	2.76631	2.98869
VE	4	8	5.3681	5.78568	7.21688	7.14705
AL	8	4	5.94444	6.4733	7.73771	7.88999
	16	2	7.04848	6.74828	8.02983	8.32372
	32	1	6.69432	6.31722	OOM <sup>(1)</sup>	7.97788
	1	1	12.7187	17.2333	22.9948	17.2444
	2	1	15.2734	18.4187	24.5339	20.31
	4	1	15.9005	18.9973	24.5516	21.899
	6	1	17.0141	19.2494	OOM <sup>(1)</sup>	22.3401
XK	8	1	16.0738	OOM <sup>(1)</sup>	OOM <sup>(1)</sup>	22.3905
	10	1	17.1923	OOM <sup>(1)</sup>	OOM <sup>(1)</sup>	22.7505
	12	1	17.3912	OOM <sup>(1)</sup>	OOM <sup>(1)</sup>	22.5492
	14	1	17.2224	OOM <sup>(1)</sup>	OOM <sup>(1)</sup>	22.7725
	16	1	17.4242	OOM <sup>(1)</sup>	OOM <sup>(1)</sup>	OOM <sup>(1)</sup>

(1) OOM : Out of Memory



Figure 7. HPCG performance of HPCG-MPMD-CPU and HPCG-MPMD-GPU on a single XE or XK node



Figure 8. HPCG performance per MPI rank of HPCG-MPMD-CPU and HPCG-MPMD-GPU on a single XE or XK node

Before moving on MPMD runs, we performed scalability tests for HPCG-MPMD-CPU and HPCG-MPMD-GPU with  $128 \times 64^2$  local block per MPI rank, as presented in Table IX. We considered two different combinations of MPI rank and thread on XE and three different MPI ranks per GPU on XK. In Figure 9, all of the cases show good scalability parallel to ideal linear scaling (i.e., orange dotted line). All cases on XK nodes performed similarly to each other except that results with 12 MPI/GPU on 16 and 32 XK nodes were relatively lower than others. On XE nodes, the case with 2 thread/MPI was 14 to 20% faster than the case with 8 thread/MPI. Figure 9 also shows that a XK node is approximately 4 times faster than an XE node, consistent with the ratios of XK to XE peak FLOPs and memory bandwidths.

In order to minimize load imbalance between XE and XK nodes during MPMD runs, we want to synchronize per-MPI performance on XE nodes with that on XK nodes. Table X and Figure 10 show HPCG performance per MPI rank of HPCG-MPMD-CPU and HPCG-MPMD-GPU. The per-MPI performance of 8 thread/MPI on XE nodes (i.e., 1.6 to 1.8 GFLOP/s) is much closer to cases of 10MPI/GPU (i.e., 2.0 to 2.3 GFLOP/s) and 12 MPI/GPU (i.e., 1.4 to 1.9 GFLOP/s) on XK nodes than the per-MPI performance of 2 thread/MPI on XE nodes (i.e., 0.45 to 0.52 GFLOP/s); therefore, we expect that 8 thread/MPI on XE nodes would perform more optimally for MPMD runs than 2 thread/MPI on XE nodes. Since the per-MPI performance of 12 MPI/GPU on XK nodes (i.e., 2.4 to 2.8 GFLOP/s) is too high to be synchronized with 8 thread/MPI on XE nodes, we did not consider it for further HPCG benchmarking on MPMD mode.

TABLE IX. HPCG PERFORMANCE OF HPCG-MPMD-CPU AND HPCG-MPMD-GPU WITH UP TO 128 XE OR XK NODES(UNIT: GFLOP/S)

Number	HPCG-HPMD-CPU		HPCG-MPMD-GPU			
of nodes	2thr/mpi <sup>(1)</sup>	8thr/mpi	8mpi/gpu <sup>(2)</sup>	10mpi/gpu	12mpi/gpu	
1	8.31327	7.14167	22.2208	22.5616	22.3774	
2	16.2669	14.1543	43.985	44.5186	44.4957	
4	32.4941	28.0463	87.3184	88.2059	86.0493	
8	65.1978	54.7415	170.352	175.311	170.902	
16	124.731	108.783	339.408	338.977	282.692	
32	248.387	216.612	648.375	692.816	525.783	
64	493.589	409.541	1266.32	1273.18	1173.86	
128	930.88	812.448	2498.76	2529.21	2245.37	

(1) thr/mpi: number of threads per MPI rank, (2) mpi/gpu: number of MPI ranks per GPU



Figure 9. HPCG performance of HPCG-MPMD-CPU and HPCG-MPMD-GPU with up to 128 XE or XK nodes

TABLE X.	HPCG PERFORMANCE PER MPI OF HPCG-MPMD-CPU
ND HPCG-MPM	ID-GPU WITH UP TO 128 XE/XK NODES(UNIT: GFLOP/S)

Number	HPCG-HP	MD-CPU	HPCG-MPMD-GPU			
of nodes	2thr/mpi <sup>(1)</sup>	8thr/mpi	8mpi/gpu <sup>(2)</sup>	10mpi/gpu	12mpi/gpu	
1	0.5196	1.7854	2.7776	2.2562	1.8648	
2	0.5083	1.7693	2.7491	2.2259	1.8540	
4	0.5077	1.7529	2.7287	2.2051	1.7927	
8	0.5094	1.7107	2.6618	2.1914	1.7802	
16	0.4872	1.6997	2.6516	2.1186	1.4724	
32	0.4851	1.6923	2.5327	2.1651	1.3692	
64	0.4820	1.5998	2.4733	1.9893	1.5285	
128	0.4545	1.5868	2.4402	1.9759	1.4618	

(1) thr/mpi: number of threads per MPI rank, (2) mpi/gpu: number of MPI ranks per GPU



Figure 10. HPCG performance per MPI rank of HPCG-MPMD-CPU and HPCG-MPMD-GPU with up to 128 XE or XK nodes

We considered four cases for HPCG performance test on MPMD mode with up to 128 XE nodes and 128 XK nodes. Case I was tested with 8 thread/MPI on XE and 10 MPI/GPU on XK. In Cases II and III, we assigned 2 thread/MPI on XE with 10 MPI/GPU on XK and 12 MPI/GPU on XK, respectively. As a reference, the case IV used only CPUs on XE (i.e., 2 CPUs/node) and XK nodes (i.e., 1 CPU/node) with 2 thread/MPI. In Table XI and Figure 11, case I shows the best HPCG performance that is around 60% to 100% faster than other cases.

Table XII shows their per-MPI performance in GFLOP/s. The per-MPI performance of case I is similar to the per-MPI performance of 8 thread/MPI on XE nodes in Table X; besides, the per-MPI performance of cases II, III and IV is similar to the per-MPI performance of 2 thread/MPI on XE nodes in Table X. Since the per-MPI performance of GPUs was better than the per-MPI performance of CPUs in all cases, the overall performance was always bounded by the CPU performance. Due to the limited size of GPU memory, we could not assign more MPI rank per GPU in order to lower the per-MPI performance on XK nodes. In addition, we could not assign more threads per MPI ranks on XE node, because of the limited threading performance between NUMA cores on XE nodes. The followings are summary of HPCG performance on the MPMD mode:

- The overall HPCG performance on the MPMD mode is determined by the minimum per-MPI performance on XE and XK nodes.
- The HPCG performance on XK nodes is around 2.7 times higher than that on XE nodes.
- In order to decrease the per-MPI performance on XK nodes, multiple MPI ranks is assigned to GPUs, but the number of MPI ranks is bounded by the memory size of GPUs (i.e., 6 GB/GPU).
- In order to improve the per-MPI performance on XE nodes, multiple threads is assigned to CPUs; however, the maximum number of threads is

bounded by the number of integer cores in NUMA core on XE nodes (i.e., 8 threads/MPI)

• As a result, the MPMD run with the combination of 8 thread/MPI on XE and 10 MPI/GPU on XK shows an optimal performance on Blue Waters.

XE nodes	XK nodes	Case I <sup>(1)</sup>	Case II <sup>(2)</sup>	Case III <sup>(3)</sup>	Case IV <sup>(4)</sup>
1	1	24.0831		14.2045	12.2138
2	2	46.7258	26.6157	28.2871	24.3066
4	4	92.937	53.0355	55.8481	48.7609
8	8	182.796	105.505	106.092	93.1538
16	16	350.558	201.75	208.912	159.893
32	32	693.168	402.937	417.709	371.003
64	64	1366.53	803.817	815.194	725.243
128	128	2599.56	1571.16	1609.91	1461.7

 TABLE XI.
 HPCG PERFORMANCE OF MPMD RUNS WITH UP TO 128

 XE AND 128 XK NODES (UNIT: GFLOP/S)

(1) 8 thread/MPI @ XE + 10 MPI/GPU @ XK, (2) 2 thread/MPI @ XE + 10 MPI/GPU @ XK, (3) 2 thread/MPI @ XE + 12 MPI/GPU @ XK, (4) 2 thread/MPI @ XE and XK w/o GPU Gray cell: HPCG could not initialize the global domain with the given number of MPI ranks.



Figure 11. HPCG performance of MPMD runs with up to 128 XE nodes and 128 XK nodes

TABLE XII. HPCG PERFORMANCE PER MPI RANK OF MPMD RUNS WITH UP TO 128 XE AND 128 XK NODES (UNIT: GFLOP/S)

XE nodes	XK nodes	Case I <sup>(1)</sup>	Case II <sup>(2)</sup>	Case III <sup>(3)</sup>	Case IV <sup>(4)</sup>
1	1	1.7202		0.5073	0.5089
2	2	1.6688	0.5118	0.5051	0.5064
4	4	1.6596	0.5100	0.4986	0.5079
8	8	1.6321	0.5072	0.4736	0.4852
16	16	1.5650	0.4850	0.4663	0.4164
32	32	1.5473	0.4843	0.4662	0.4831
64	64	1.5251	0.4831	0.4549	0.4722
128	128	1.4506	0.4721	0.4492	0.4758

8 thread/MPI @ XE + 10 MPI/GPU @ XK, (2) 2 thread/MPI @ XE + 10 MPI/GPU @ XK,
 3) 2 thread/MPI @ XE + 12 MPI/GPU @ XK, (4) 2 thread/MPI @ XE and XK w/o GPU Gray cell: HPCG could not initialize the global domain with the given number of MPI ranks.

### III. HPGMG BENCHMARK ON BLUE WATERS

### A. General Description

HPGMG [4,5] is another effort for HPC performance benchmarking based on geometric multi-grid methods with emphasis on community-driven development process, longterm durability, scale-free specification and scale-free communication. It provides two implementations, Finite Element (HPGMG-FE) and Finite Volume (HPGMG-FV) implementations; HPGMG-FE is compute-intensive and cache-intensive, while HPGMG-FV is memory bandwidthintensive. HPGMG-FV has been used for the official list, and it solves an elliptic problem on isotropic Cartesian grids with fourth-order accuracy in the max norm. It calculates a flux term on each of the 6 faces on every cell in the entire domain. The fourth-order implementation [5] requires 4 times the floating-point operations, 3 times the MPI messages per smoother and 2 times the MPI message size without additional DRAM data movement compared to the second-order implementation [4] proposed originally. The solution process employs the Full Multi-grid (FMG) F-cycle that is a series of progressively deeper geometric multi-grid V-cycles. There are several dozen stencils that vary in shape

and size, sweep per step, and the grid sizes vary exponentially. Coarse grid solution process can occur on a single core of a single node, and then the coarse grid solution is propagated to every thread in the system. At SC16, the 5<sup>th</sup> HPGMG-FV performance list was announced and it has two GPU-enabled systems. In the list, those GPU-enabled systems provide GPU-based and CPU-based performance separately.

In this section, we first performed HPGMG tests on CPU-based or GPU-enabled nodes on Blue Waters, and then analyzed several major components that affected the performance. We then conducted MPMD runs on CPUbased and GPU-enabled nodes together for HPGMG. We provide our analyses and recommendations for other geometric multi-grid applications on hybrid HPC platforms.

# B. HPGMG on CPU-based XE nodes

We downloaded HPGMG-FV source files from the official HPGMG Bitbucket repository (i.e., https://bitbucket. org/hpgmg/hpgmg/). We used them (i.e., commit: 8a2f0e1) to build a HPGMG binary for CPU-based XE nodes (i.e., HPGMG-CPU-8a2f0e1) with gcc/4.9.3 and cray-mpich/7.3.3 on Blue Waters. HPGMG-FV requires two input parameters

nthu	ь	nB										
num	к	1	2	3	4	5	6	7	8	16	32	wiax
	4	0.0820	0.0956	0.0596	0.1489	0.1514	0.1367	0.1369	0.2007			
	5	0.2435	0.2517	0.1977	0.2931	0.2924	0.2864	0.2873	0.3320			
1	6	0.3788	0.3731	0.3396	0.3741	0.3730	0.3826	0.3833	0.4127			0 4756
1	7	0.4445	0.4344	0.4044	0.4131	0.4131	0.4293	0.4297	0.4588			0.4/56
	8	0.4756	OOM <sup>(1)</sup>	OOM	OOM	OOM	OOM	OOM	OOM			
	9	OOM	OOM	OOM	OOM	OOM	OOM	OOM	OOM			
	4	0.0453	0.0997	0.0696	0.1111	0.1111	0.0823	0.1608	0.1606			
	5	0.1623	0.2664	0.2184	0.2712	0.2711	0.2341	0.3134	0.3147			
2	6	0.3109	0.4005	0.3670	0.3914	0.3915	0.3634	0.4081	0.4083			0.4804
2	7	0.3768	0.4508	0.4269	0.4412	0.4410	0.4129	0.4490	0.4488			0.4804
	8	0.4041	0.4804	OOM	OOM	OOM	OOM	OOM	OOM			
	9	OOM	OOM	OOM	OOM	OOM	OOM	OOM	OOM			
	4	0.0216	0.0298	0.0414	0.0726	0.0739	0.0540	0.0541	0.0872			0.4496
	5	0.0975	0.1216	0.1505	0.2177	0.2177	0.1817	0.1813	0.2322			
	6	0.2633	0.2680	0.2979	0.3653	0.3649	0.3347	0.3345	0.3613			
4	7	0.3893	0.3503	0.3719	0.4271	0.4270	0.4039	0.4040	0.4185			
	8	0.4288	0.3765	0.3978	0.4493	0.4496	OOM	OOM	OOM			
	9	OOM	OOM	OOM	OOM	OOM	OOM	OOM	OOM			
	4	0.0141	0.0159	0.0160	0.0214	0.0214	0.0346	0.0346	0.0581			
	5	0.0634	0.0733	0.0733	0.0919	0.0921	0.1278	0.1279	0.1826			
<u>ه</u>	6	0.2103	0.2237	0.2238	0.2377	0.2377	0.2776	0.2778	0.3416			0.4494
0	7	0.3770	0.3728	0.3728	0.3444	0.3443	0.3720	0.3720	0.4267			0.4494
	8	0.4346	0.4230	0.4232	0.3744	0.3745	0.3984	0.3983	0.4494			
	9	0.4352	OOM	OOM	OOM	OOM	OOM	OOM	OOM			
	4	0.0028	0.0081	0.0081	0.0103	0.0102	0.0103	0.0137	0.0137			
	5	0.0147	0.0380	0.0379	0.0486	0.0485	0.0484	0.0633	0.0637			
16	6	0.0701	0.1319	0.1318	0.1465	0.1463	0.1463	0.1695	0.1694			0 3 3 3 5
10	7	0.1873	0.2597	0.2600	0.2615	0.2615	0.2614	0.2733	0.2732	0.3003	0.3006	0.5555
	8	0.2644	0.3239	0.3241	0.3160	0.3161	0.3161	0.3167	0.3168	0.3335	OOM	
	9	0.2706	0.3219	0.3221	OOM	OOM	OOM	OOM	OOM			
	4	0.0012	0.0015	0.0015	0.0060	0.0051	0.0051	0.0051	0.0064			
	5	0.0056	0.0080	0.0081	0.0244	0.0245	0.0244	0.0244	0.0317			
32	6	0.0258	0.0410	0.0411	0.0920	0.0921	0.0922	0.0921	0.1045			0.2405
32	7	0.0732	0.1194	0.1196	0.1772	0.1774	0.1773	0.1773	0.1771	0.1864	0.2072	0.2493
	8	0.1132	0.1877	0.1875	0.2387	0.2386	0.2386	0.2387	0.2342	0.2371	0.2495	
	9	0.1146	0.1934	0.1932	0.2358	0.2357	0.2359	0.2357	OOM			

 TABLE XIII.
 HPGMG PERFORMANCE ON 16 XE NODES (UNIT: GDOF/S)

(1) OOM: Out-of-Memory, Configurations in gray cells were not tested.

(i.e., k and nB at the following bullets) to construct the global cubic geometry and distribute local boxes to each MPI ranks. In addition, number of thread per MPI rank (i.e., *nthr* at the following bullets) needs to be set to concurrently solve assigned local boxes on MPI ranks using the MPI-OpenMP hybrid implementation. The followings are notations for three parameters with their brief definitions:

- *k*: the log base 2 of the dimension of each box on the finest grid
- *nB*: target number of boxes per process: a loose bound on memory per process.
- *nthr*: number of thread for OpenMP

At first, we tested HPGMG-CPU-8a2f0e1 on 16 XE nodes for a large number of sets for three parameters (i.e., k= 4, 5, 6, 7, 8, 9, and nB = 1, 2, 3, 4, 5, 6, 7, 8, 16, 32, andnthr = 1, 2, 4, 8, 16, 32, and Table XIII presents HPGMG results for the finest resolution (i.e., 1h). Figure 12 shows plots for HPGMG performance vs. number of thread per MPI rank (i.e., *nthr*). Blue circles in Figure 12 expresses the maximum HPGMG performance for each nthr. They are very close to a red line in the figure that represents HPGMG results with 256<sup>3</sup> boxes per thread (i.e., k=8 with *nthr=nB*). In the figure, there are four additional lines for 128<sup>3</sup> boxes per thread (i.e., k=7 with *nthr=nB*), 64<sup>3</sup> boxes per thread (i.e., k=6 with *nthr=nB*),  $32^3$  boxes per thread (i.e., k=5 with *nthr=nB*), and  $16^3$  boxes per thread (i.e., *k*=4 with *nthr=nB*). In the cases for 256<sup>3</sup> and 128<sup>3</sup> boxes per thread, we tested them with 16 thread/MPI rank and 32 thread/MPI rank. We observed significant performance loss on those cases due to inefficiencies in OpenMP implementations over NUMA nodes (i.e., 16 thread/MPI) or CPU sockets (i.e., 32 thread /MPI). We note that HPGMG performance improves as the number of boxes per thread increases (see Figure 13). This is understandable because, as local box sizes grow, local calculations for fine grid solutions become enriched and communication costs for coarse grid solutions can rapidly drop. Figure 14 shows plots for HPGMG performance vs. target number of boxes per MPI rank (i.e., nB). Depending on the ratio of actually distributed boxes per MPI rank that is usually less than nB, HPGMG performance fluctuates a little bit for the same size of local boxes; however, the performance difference between  $128^3$  and  $64^3$  local boxes is more obvious than the oscillation of the performance with the same size of local boxes.

With 256<sup>3</sup> boxes per thread, we performed scalability tests with up to 36% Blue Waters XE nodes (i.e., 2, 16, 128, 1024 and 8192 XE nodes). Table XIV presents selected input parameters (i.e., *nthr*, *k* and *nB*), number of employed XE nodes, number of MPI ranks, number of boxes assigned to each thread, total number of boxes, HPGMG performance for three resolutions (i.e., 1h, 2h and 4h), and parallel efficiency based on 2 XE-node performance of each *nthr*. On 8192 XE nodes, HPGMG-CPU-8a2f0e1 shows the best performance with 1 thread/MPI rank, and its parallel efficiency is 93% that is very impressive. In other cases, parallel efficiency is more than or equal to 87% on more than



Figure 12. HPGMG performance vs. number of threads/MPI-rank of HPGMG-CPU-8a2f0e1 on 16 XE nodes



Figure 13. HPGMG performance vs. number of boxes/thread of HPGMG-CPU-8a2f0e1 on 16 XE nodes



Figure 14. HPGMG performance vs. target number of boxes/MPI rank of HPGMG-CPU-8a2f0e1 on 16 XE nodes

	Ŀ	D	XE	Number of	Number of cells per	Number of	HPC	GMG (GDO	F/s)	Parallel	HPGMG
nınr	ĸ	пБ	nodes	MPI ranks	thread	cells in global	1h	2h	4h	efficiency (%)	/MPI rank
1	8	1	2	64	256 <sup>3</sup>	1024 <sup>3</sup>	0.06016	0.05753	0.04387	100	0.000940
1	8	1	16	512	256 <sup>3</sup>	2048 <sup>3</sup>	0.474	0.4441	0.3767	98	0.000926
1	8	1	128	4096	256 <sup>3</sup>	4096 <sup>3</sup>	3.762	3.466	2.752	98	0.000918
1	8	1	1024	32768	256 <sup>3</sup>	8192 <sup>3</sup>	29.35	26.18	18.62	95	0.000896
1	8	1	8192	262144	256 <sup>3</sup>	16384 <sup>3</sup>	229.6	198	126.3	93	0.000876
2	8	2	2	32	256 <sup>3</sup>	1024 <sup>3</sup>	0.06054	0.05777	0.05312	100	0.001892
2	8	2	16	256	256 <sup>3</sup>	2048 <sup>3</sup>	0.4708	0.4493	0.3952	97	0.001839
2	8	2	128	2048	256 <sup>3</sup>	4096 <sup>3</sup>	3.784	3.496	2.884	98	0.001848
2	8	2	1024	16384	256 <sup>3</sup>	8192 <sup>3</sup>	28.85	25.39	19.03	93	0.001761
2	8	2	8192	131072	256 <sup>3</sup>	16384 <sup>3</sup>	223.7	189.5	125.1	90	0.001707
4	8	4	2	16	256 <sup>3</sup>	1024 <sup>3</sup>	0.05672	0.05449	0.04811	100	0.003545
4	8	4	16	128	256 <sup>3</sup>	2048 <sup>3</sup>	0.4473	0.4245	0.3591	99	0.003495
4	8	4	128	1024	256 <sup>3</sup>	4096 <sup>3</sup>	3.526	3.286	2.654	97	0.003443
4	8	4	1024	8192	256 <sup>3</sup>	8192 <sup>3</sup>	26.63	23.3	17.08	92	0.003251
4	8	4	8192	65536	256 <sup>3</sup>	16384 <sup>3</sup>	205.9	172.8	112.5	89	0.003142
8	8	8	2	8	256 <sup>3</sup>	1024 <sup>3</sup>	0.05662	0.05462	0.04338	100	0.007078
8	8	8	16	64	256 <sup>3</sup>	2048 <sup>3</sup>	0.4485	0.4244	0.3367	99	0.007008
8	8	8	128	512	256 <sup>3</sup>	4096 <sup>3</sup>	3.503	2.312	2.463	97	0.006842
8	8	8	1024	4096	256 <sup>3</sup>	8192 <sup>3</sup>	26.3	22.86	15.59	91	0.006421
8	8	8	8192	32768	256 <sup>3</sup>	16384 <sup>3</sup>	201.5	165.7	100.4	87	0.006149

TABLE XIV. HPGMG PERFORMANCE OF HPGMG-CPU-8A2F0E1 WITH UP TO 36% OF BLUE WATERS XE NODES (UNIT: GDOF/S)



Figure 15. HPGMG performance for various number of thread per MPI rank with up to 36% of Blue Waters XE nodes

one third of Blue Waters XE nodes. The HPGMG performance per MPI rank improves as the number of threads increases, and its range is between 0.9 and 7.1 MDOF/s. Figure 15 illustrates HPGMG performance for the finest resolution (i.e., 1h). Four solid lines show HPGMG performance with  $256^3$  boxes per thread for 1 thread/MPI rank (i.e., blue line), 2 thread/MPI rank (i.e., red line), 4 thread/MPI rank (i.e., green line) and 8 thread/MPI rank (i.e., purple line). Light blue dotted line represents an ideal linear scaling. In most cases, blue or red line shows the best performance. Green and purple lines are lower than blue or read lines but the difference between the best and the worst cases is less than 11 %. Figure 16 shows HPGMG performance for different resolutions with 1 thread/MPI rank. Blue line is for the finest resolution (i.e., 1h), while red and green lines are for 2h and 4h, respectively. Purple dotted line shows an ideal linear scaling. Blue and red lines for 1h



Figure 16. HPGMG performance in terms of resolutions (i.e., 1h, 2h and 4h) with up to 36% of Blue Waters XE nodes

and 2h are almost parallel to the purple dotted line. The green line for 4h is sub-linear.

#### C. HPGMG on GPU-enabled XK nodes

NVIDIA has shared HPGMG with GPU acceleration via Bitbucket (i.e., https://bitbucket. org/nsakharnykh/hpgmgcuda). It is based on HPGMG v0.3 compliant MPI version 2, OpenMP version 3 and CUDA version 7. Among multi-grid F-cycles, top levels with fine resolutions run on GPUs, while bottom levels with coarse resolutions run on host CPUs that has less latency than GPUs. We downloaded HPGMG-CUDA source file (commit: 02c7ea214ce8) and built an HPGMG-CUDA binary (i.e., HPGMG-GPU-02c7ea2) with CUDA 7.5 and cray-mpich/7.3.3 on Blue Waters.

nMPI	Ŀ	nB	XK	nMPI <sup>(2)</sup>	Boxes	Total	HPGMG	
/GPU <sup>(1)</sup>	к	пр	nodes	IIIVIFI	/MPI <sup>(3)</sup>	boxes	(1h)	
1	8	1	16	16	1	512 <sup>3</sup>	0.3304	
1	8	2	16	16	1.688	768 <sup>3</sup>	0.5639	
1	8	3	16	16	1.688	768 <sup>3</sup>	0.5636	
2	8	1	16	32	1.688	768 <sup>3</sup>	0.589	
3	8	1	16	48	0.562	768 <sup>3</sup>	0.5909	
1	7	8	16	16	7.812	640 <sup>3</sup>	0.545	
1	7	9	16	16	7.812	640 <sup>3</sup>	0.5447	
1	7	14	16	16	13.5	768 <sup>3</sup>	0.6195	
1	7	16	16	16	13.5	768 <sup>3</sup>	0.6194	
7	7	2	16	112	1.929	768 <sup>3</sup>	0.5837	
7	7	3	16	112	1.929	768 <sup>3</sup>	0.5829	
1	6	112	16	16	108	768 <sup>3</sup>	0.6433	
1	6	108	16	16	108	768 <sup>3</sup>	0.6433	
2	6	54	16	32	54	768 <sup>3</sup>	0.6442	
4	6	27	16	64	27	768 <sup>3</sup>	0.646	
8	6	14	16	128	13.5	768 <sup>3</sup>	0.5937	

 
 TABLE XV.
 HPGMG performance of HPGMG-GPU-02c7ea2 on 16 XK nodes (unit: GDOF/s)

(1) Number of MPI rank assigned to each GPU

(2) Number of total MPI rank. (3) Number of boxes per MPI rank



Figure 17. Performance of HPGMG-GPU-02c7ea2 on 16 XK nodes

In this section, we consider an additional parameter, number of MPI rank per GPU (i.e., *nMPI/GPU*) that affects the HPGMG-CUDA performance. Table XV and Figure 17 present the performance of HPGMG-GPU-02c7ea2 on 16

XK nodes for several sets of parameters (i.e., nMPI/GPU, k and nB). Similarly to our discussion in the previous section, HPGMG performance on XK nodes is proportional to total number of boxes. The sets with k=6 show the best HPGMG performance on 16 XK nodes.

We performed scalability tests for HPGMG-GPU-02c7ea2 with up to 24% Blue Waters XK nodes with the most optimal configurations (i.e., k=6, nMPI/GPU = 1, 2, 4and nB = 108, 54, 27) from the tests on 16 XK nodes. While cases with nMPI/GPU=2 and 4 on 1024 XK nodes failed to create levels due to "malloc failed" error, other cases ran smoothly, as presented in Table XVI and Figure 18. Parallel efficiency drops very rapidly as number of XK nodes increases. In the worst case, it is 63% on 1024 XK nodes, which is much lower than the worst case on 8192 XE nodes (i.e., 87% in Table XIV). The blue dotted line in Figure 18 represents an ideal linear scaling, and all cases in Figure 18 show sub-linear scaling. We note that the HPGMG performance drops between different resolutions (i.e., 1h vs. 2h vs. 4h) are significantly large compared to our XE scalability tests (see Figure 16). It looks smaller local blocks on XK nodes than on XE nodes results in the huge performance drops. It follows the relation of HPGMG performance and number of boxes per thread shown in Figure 13. HPGMG performance per MPI rank decreases as *nMPI/GPU* increases. It ranges from 9.7 to 32 MDOF/s.



Figure 18. HPGMG performance of HPGMG-GPU-02c7ea2 with up to 24% Blue Waters XK nodes

nMPI/	MPI/ k nB		XK	Number of	Number of cells per	Number of	HPC	GMG (GDO	F/s)	Parallel	HPGMG
GPU <sup>(1)</sup>	к	nD	nodes	MPI ranks	MPI ranks	cells in global	1h	2h	4h	efficiency (%)	/MPI rank
1	6	108	2	2	$108 \times 64^{3}$	384 <sup>3</sup>	0.08475	0.04496	0.0144	100%	0.042375
1	6	108	16	16	$108 \times 64^{3}$	768 <sup>3</sup>	0.6433	0.3325	0.1043	95%	0.040206
1	6	108	128	128	$108 \times 64^{3}$	1536 <sup>3</sup>	4.611	2.271	0.697	85%	0.036023
1	6	108	1024	1024	$108 \times 64^{3}$	3072 <sup>3</sup>	27.24	13.11	4.104	63%	0.026602
2	6	54	2	4	$54 \times 64^3$	384 <sup>3</sup>	0.08539	0.0486	0.01677	100%	0.021348
2	6	54	16	32	$54 \times 64^3$	768 <sup>3</sup>	0.6446	0.3463	0.1133	94%	0.020144
2	6	54	128	256	$54 \times 64^3$	1536 <sup>3</sup>	4.734	2.419	0.7401	87%	0.018492
2	6	54	1024	2048	$54 \times 64^3$	3072 <sup>3</sup>		malloc fail	ed - create_	level/level->my_bo	xes
4	6	27	2	8	$27 \times 64^{3}$	384 <sup>3</sup>	0.08531	0.05022	0.01801	100%	0.010664
4	6	27	16	64	$27 \times 64^{3}$	768 <sup>3</sup>	0.6492	0.366	0.1217	95%	0.010144
4	6	27	128	512	$27 \times 64^{3}$	1536 <sup>3</sup>	4.972	2.675	0.8368	91%	0.009711
4	6	27	1024	4096	$27 \times 64^{3}$	$3072^{3}$		malloc fail	ed - create	level/level->mv bo	xes

TABLE XVI. HPGMG PERFORMANCE OF HPGMG-GPU-02C7EA2 WITH UP TO 24% OF BLUE WATERS XK NODES (UNIT: GDOF/S)

(1) Number of MPI rank assigned to each GPU

# D. HPGMG on CPU-based XE and GPU-enabled XK nodes on MPMD mode

MPMD mode requires the same base code for CPUbased and GPU-enabled executable binaries; therefore, we downloaded HPGMG v0.3 (commit: 26f74a2120ce) that is an older version than HPGMC-CPU-8a2f0e1. We then built an additional CPU based binary (i.e., HPGMG-CPU-26f74a2) for our MPMD runs. As we discussed in the previous section for HPCG MPMD mode tests, we wanted to synchronize HPGMG performance per MPI rank on XE and XK nodes. For that, we needed to increase nthr (number of thread) to improve the per-MPI performance of CPU-based binary. We also increased nMPI/GPU (number of MPI rank per GPU) in order to reduce the per-MPI performance of GPU-enabled binary. Since the GPU memory per XK node (i.e., 6 GB) was much smaller than CPU memory per XE node (i.e., 64 GB), we also reduced the box size assigned to each XE node to the level of each XK node. After several parametric tests, we selected a set of parameters (i.e., *nthr*=8, *nMPI/GPU=*7, k=7 and nB=2) for an optimal MPMD run by

synchronizing per-MPI performances on XE and XK nodes. As references, we selected two sets of parameters; one is for the best CPU-based binary (i.e., nthr=1, k=8 and nB=1) and the other is for the best GPU-enable binary (i.e., nMPI/GPU=1, k=6 and nB=108).

Table XVII, Figures 19 and 20 show HPGMG performance on MPMD mode, XE-only mode and XK-only mode. We tested with 16 XE and 16 XK nodes, and then with 128 XE and 128 XK nodes. Unfortunately, one of the reference sets for the best CPU-based binary (i.e., *nthr*=1, k=8 and nB=1) encountered out-of-memory errors. It seems the recent HPGMG (commit: 8a2f0e1123c7) is better at the memory usage for the same size of problems than HPGMG v0.3 (commit: 26f74a2120ce).

With the first set of parameters for the synchronized per-MPI performance (i.e., *nthr*=8, *nMPI/GPU*=7, *k*=7 and *nB*=2), it turned out HPGMG performance on MPMD mode was similar to summation of HPGMG performances of XEonly and XK-only modes. On 16 XE and 16 XK nodes, HPGMG at each MPI rank (i.e., 4.516 MDOF/s) was

TABLE XVII. BOXESHPGMG PERFORMANCE OF HPGMG-CPU-26F74A2 AND HPGMG-GPU-02C7EA2 ON MPMD MODE (UNIT: GDOF/S)

Dup type (with a MDI/CDI t, a P)	XE	XK	nMPI	nMPI	Number	Boxes	HPGMG (GDOF/s)			HPGMG
Kun type ( <i>ninr</i> , <i>nimr1/GPU</i> , <i>k</i> , <i>nB</i> )	nodes	nodes	$aXE^{(1)}$	$@XK^{(2)}$	of cells	/MPI <sup>(3)</sup>	1h	2h	4h	/MPI rank <sup>(4)</sup>
MPMD (8,7,7,2)	16	16	64	112	896	1.949	0.7949	0.2997	0.0768	0.004516
XE only (8,7,7,2)	16		64		640	1.953	0.3525	0.2084	0.063	0.005508
XK only (8,7,7,2)		16		112	768	1.929	0.4828	0.1578	0.03365	0.004311
MPMD (2,1,6,108)	16	16	256	16	1792	80.706	0.4002	0.3257	0.2294	0.001471
XE only (2,1,6,108)	16		256		768	85.75	0.426	0.3437	0.2363	0.001664
XK only (2,1,6,108)		16		16	1792	108	0.6433	0.3325	0.1043	0.040206
MPMD (1,1,8,1)	16	16	512	16	2048	0.97		Out	of memory	
MPMD (8,7,7,2)	128	128	512	896	1792	1.949	5.818	1.917	0.433	0.004132
XE only (8,7,7,2)	128		512		1280	1.953	2.618	1.417	0.3976	0.005113
XK only (8,7,7,2)		128		896	1536	1.929	3.557	1.026	0.2039	0.003970
MPMD (2,1,6,108)	128	128	2048	128	3584	80.706	3.171	2.539	1.69	0.001457
XE only (2,1,6,108)	128		2048		1536	85.75	3.373	2.682	1.764	0.001647
XK only (2,1,6,108)		128		128	3584	108	4.611	2.271	0.697	0.036023
MPMD (1,1,8,1)		128		128	3584	108		Out	of memory	

(1) nMPI/XE: number of MPI ranks assigned to XE nodes, (2) nMPI/XE: number of MPI ranks assigned to XK nodes,

(3) Boxes/MPI: number of boxes (2<sup>k</sup> cells) assigned to each MPI rank, (4) HPGMG/MPI rank: HPGMG performance per MPI rank





Figure 19. HPGMG performance on MPMD mode with 16 XE nodes and 16 XK nodes

Figure 20. HPGMG performance on MPMD mode with 128 XE nodes and 128 XK nodes

between those of XE-only (i.e., 5.508 MDOF/s) and XK-only (i.e., 4.311 MDOF/s) modes. The per-MPI performance of MPMD mode on 128 XE and 128 XK nodes (i.e., 4.132 MDOF/s) was also between those of XE-only (i.e., 5.113 MDOF/s) and XK-only (i.e., 3.97 MDOF/s).

HPGMG performances on MPMD mode with the second set of parameters for the best GPU-enable performance (i.e., nMPI/GPU=1, k=6 and nB=108) were much lower than those with the first set of parameters. They were even lower than HPGMG performances on XE-only or XK-only mode. It is because the performances on MPMD mode were bounded by lower per-MPI performance that was of CPU based binary with the second set of parameters. On 16 XE and 16 XK nodes, the per-MPI performance on MPMD mode (i.e., 1.471MDOF/s) was slightly lower than that on XE-only mode (i.e., 1.664 MDOF/s) and much lower than that on XK-only mode (i.e., 40.21 MDOF/s). The HPGMG performance per MPI rank on MPMD mode with 128 XE and 128 XK nodes (i.e., 1.457 MDOF/s) was also lower than that on XE-only mode (i.e., 1.647 MDOF/s) and much lower than that on XK-only mode (i.e., 36.02 MDOF/s).

#### IV. CONCLUDING REMARKS

HPCG and HPGMG benchmarking are new metrics to measure modern HPC performance. They include many featured aspects representing important science and engineering HPC applications popularly used in modern HPC community. We performed HPCG and HPGMG benchmarking tests on CPU-based XE nodes, GPU-enabled XK nodes, and both XE and XK nodes with MPMD mode on Blue Waters.

HPCG benchmarking on CPU-based XE nodes showed consistent performance for various numbers of equations per MPI process, and the parallel efficiency kept more than or equal to 86% on tested cases (up to 4.6% Blue Waters XE nodes, 1024 XE nodes). We tested HPCG on GPU-enabled XK nodes with up to 6.1% Blue Waters XE nodes (i.e., 256 XK nodes). Its performance increased exponentially with number of equations per MPI process at the beginning, and then it converged to a certain level. The parallel efficiency of HPCG on XK nodes was more than or equal to 88% that was a little bit lower than that on the same number of XE nodes (i.e., 92%). HPCG performance on XK nodes was around 2.7 times faster than HPCG on the same number of XE nodes. For an optimal performance on MPMD mode, we tried to synchronize HPCG performance per MPI process on XE and XK nodes. For that, we increased number of threads on XE nodes and increased number of MPI ranks per GPU on XK nodes. As a result, we selected a set of parameters with 128×64<sup>3</sup> local blocks per MPI process, 8 threads per MPI process on XE nodes and 10 MPI processes per GPU on XK nodes. It resulted in around 60% to 100% better performance than other reference cases that were chosen for the best CPUbased performance or the best GPU-based performance.

HPGMG performance on XE nodes rapidly increased with number of cells assigned to each compute node, and then it was converged. Its OpenMP implementation was efficient enough to keep the performance at a certain level

with threading in a NUMA core. The parallel efficiency of HPGMG on XE nodes is impressive, at more than 87% with up to 36% of the available XE nodes (i.e., 8192 XE nodes). HPGMG on GPU-enabled XK nodes was sensitive to number of cells per MPI process, since the assignable number of cells on GPU was not big enough for HPGMG performance to be converged. Its performance on 16 XK nodes was around 34% faster than HPGMG on 16 XE nodes, but the parallel efficiency on XK nodes rapidly dropped to 63% on 1024 XK nodes (i.e., 24% Blue Waters XK nodes). That was much lower than the parallel efficiency of HPGMG on 1024 XE nodes (i.e., 91%). For MPMD mode, we again tried to synchronize HPGMG performance per MPI process on XE and XK nodes. After several parametric studies, we selected an optimal set of parameters (i.e., 8 thread per MPI process on XE nodes, 7 MPI processes per GPU on XK nodes and  $2 \times 128^3$  cells per MPI process). It turned out that HPGMG on MPMD mode with the selected set of parameters shows around 100% better performance than a reference case that was optimized for the best HPGMG performance with GPU-enabled executable binary.

In summary, we performed HPCG and HPGMG benchmarking tests on Blue Waters with various configurations. We analyzed each benchmark on CPU-based nodes, GPU-enabled nodes, and both of heterogeneous nodes with MPMD mode. In both benchmarks, the performance on GPU-enabled nodes was better than those on CPU-based nodes. To obtain an optimal performance on MPMD mode, we tried to synchronize performance per MPI process on CPU-based and GPU-enable nodes. It turned out the synchronization of the per-MPI performance resulted in 60% to 100% better performance than other reference cases. We hope this study will help Cray users optimize their applications for science and engineering projects on modern hybrid HPC systems.

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