



A CUDA IMPLEMENTATION OF THE HIGH PERFORMANCE CONJUGATE GRADIENT (HPCG) BENCHMARK

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OUTLINE

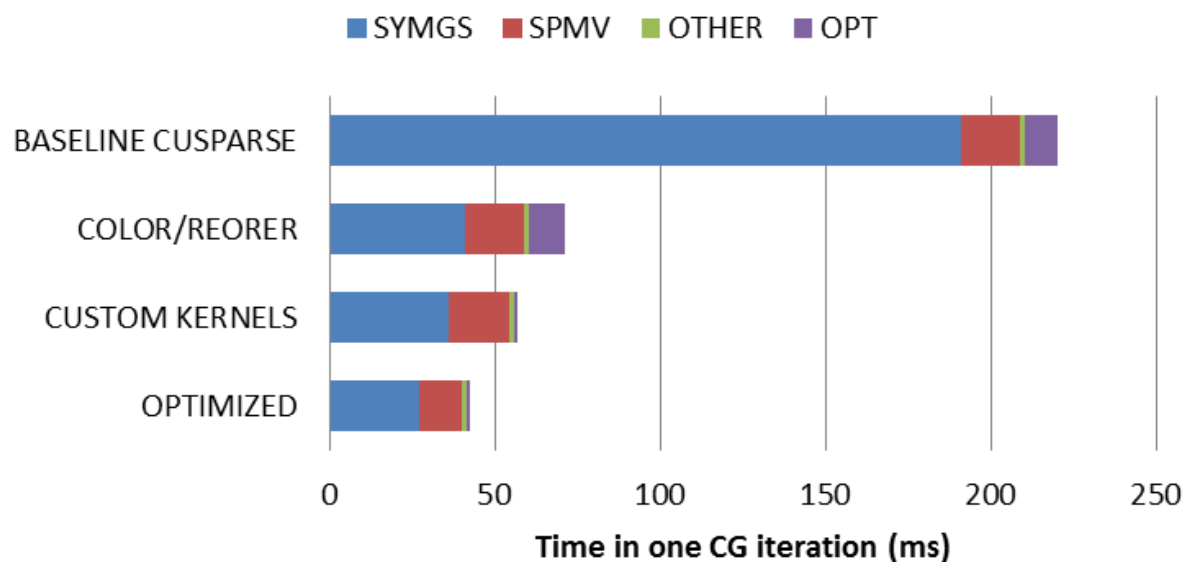
- ▶ CUDA implementation(s) overview
- ▶ Single node performance
- ▶ Multi node performance
- ▶ Comparison to other architectures
- ▶ Conclusions/suggestions

CUDA IMPLEMENTATIONS

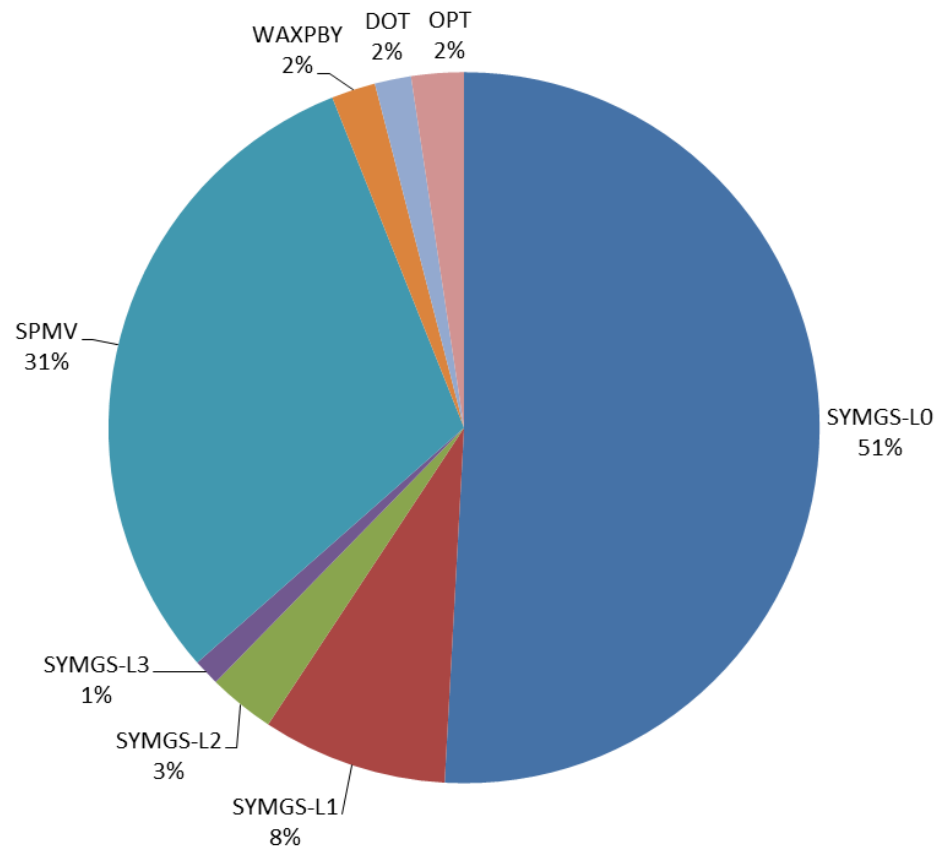
- I. Cusparsе CSR
- II. Cusparsе CSR + Matrix Reordering (graph coloring)
- III. Custom Kernels CSR + Matrix Reordering (graph coloring)
- IV. Custom Kernels ELL + Matrix Reordering (graph coloring)

RESULTS - SINGLE GPU

HPCG time comparison (K20X 128^3)

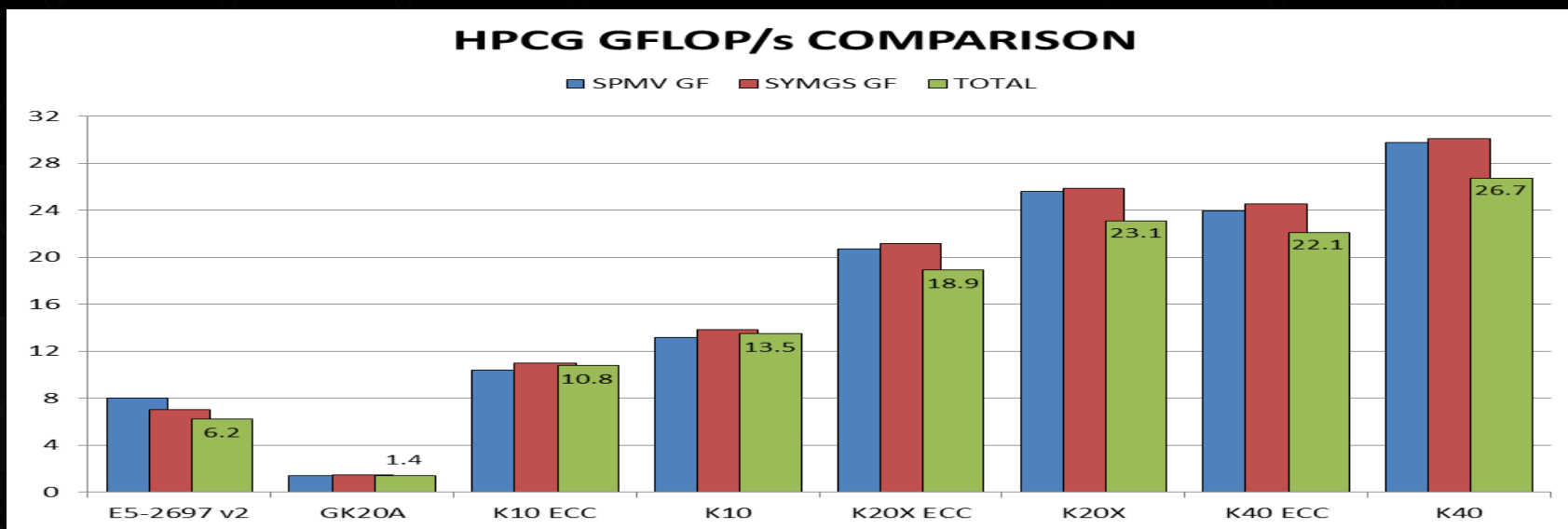


Optimized HPCG time (K20X)



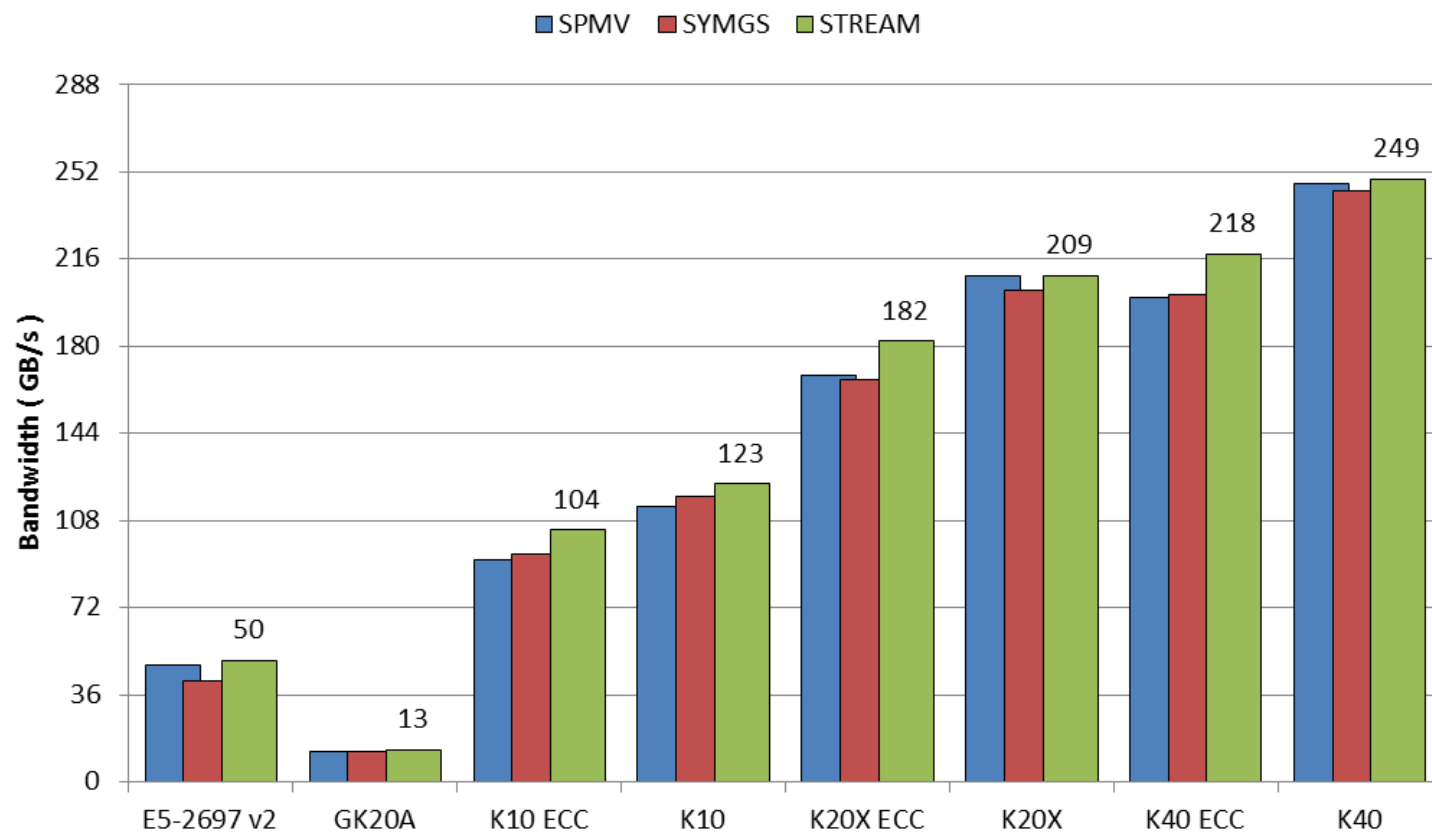
RESULTS - SINGLE GPU

GPU	#SMs	#Cores SP/DP	Core Clock	DP (Gflops)	Memory Clock	Memory Bus Width	Memory Bandwidth
Tegra K1	1	192/8	852	13.6	924	64-bit	14.7
Tesla K10	8	1536/64	745	95	2500	256-bit	160
Tesla K20x	14	2688/896	732	1310	2600	384-bit	250
Tesla K40	15	2880/960	875	1680	3000	384-bit	288

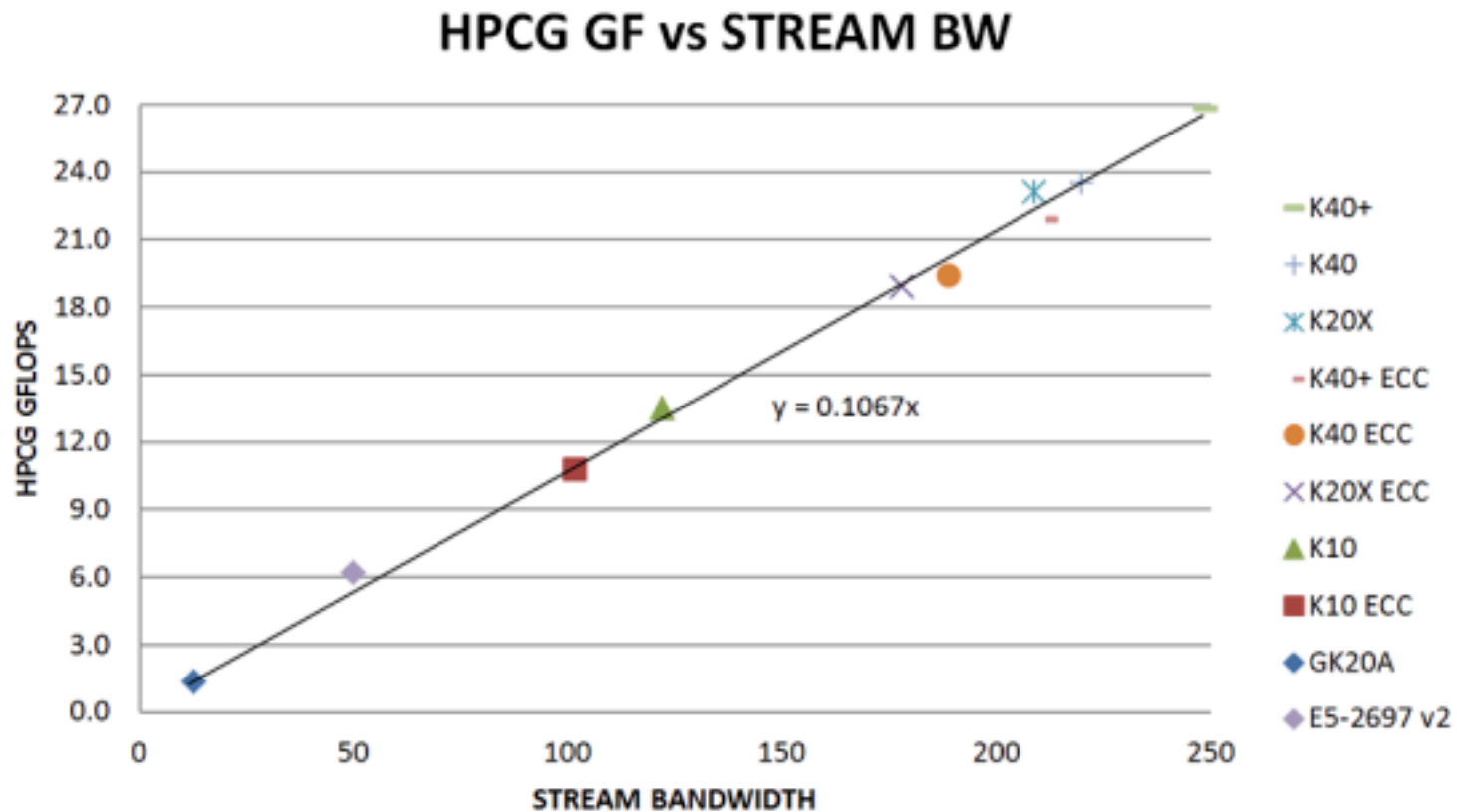


RESULTS - SINGLE GPU

HPCG BANDWIDTH COMPARISON



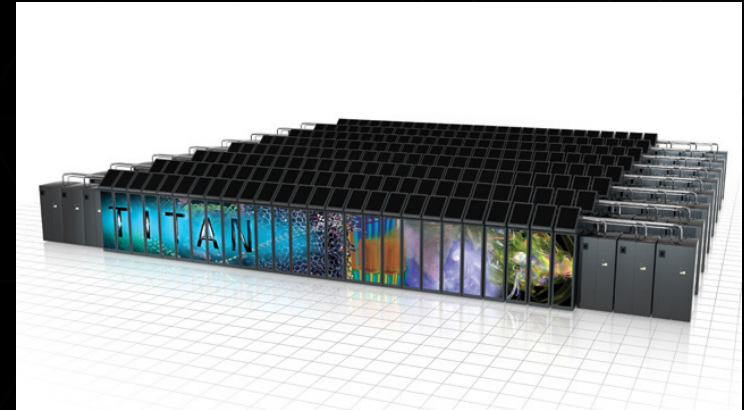
RESULTS - SINGLE GPU



RESULTS - GPU SUPERCOMPUTERS

▶ Titan @ ORNL

- ▶ Cray XK7, 18688 Nodes
- ▶ 16-core AMD Interlagos + K20X
- ▶ Gemini Network - 3D Torus Topology



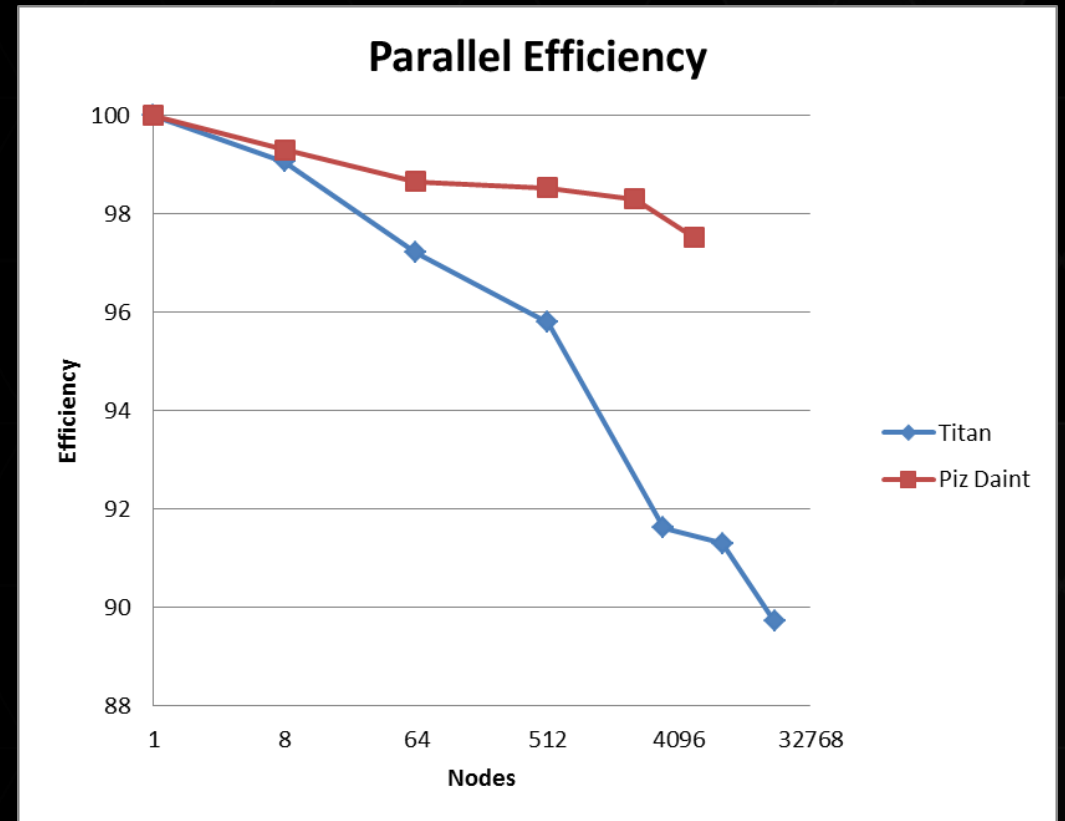
▶ Piz Daint @ CSCS

- ▶ Cray XC30, 5272 Nodes
- ▶ 8-core Xeon E5 + K20X
- ▶ Aries Network - Dragonfly Topology



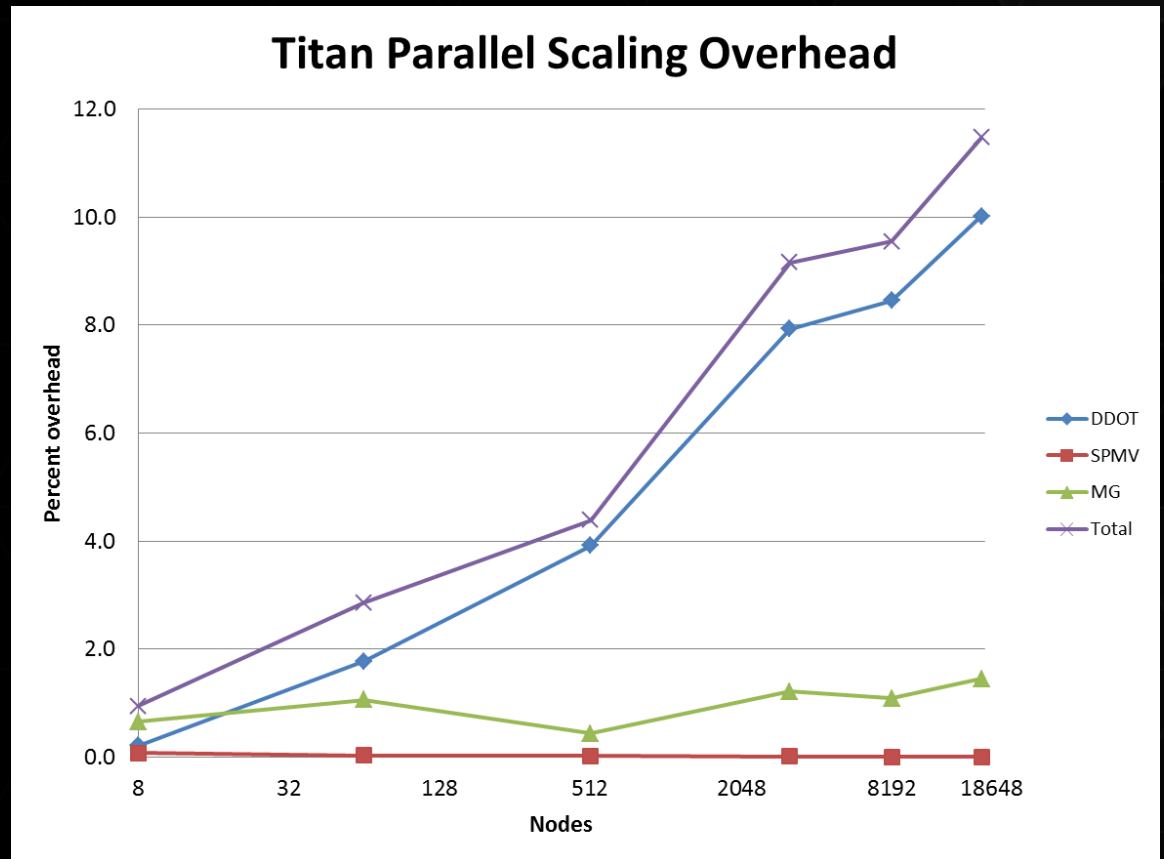
RESULTS - GPU SUPERCOMPUTERS

- ▶ 1 GPU = 20.8 GFLOPS (ECC ON)
- ▶ ~7% iteration overhead at scale
- ▶ Titan @ ORNL
 - ▶ 322 TFLOPS (18648 K20X)
 - ▶ 89% efficiency (17.3 GF per GPU)
- ▶ Piz Daint @ CSCS
 - ▶ 97 TFLOPS (5265 K20X)
 - ▶ 97% efficiency (19.0 GF per GPU)



RESULTS - GPU SUPERCOMPUTERS

- ▶ DDOT (-10%)
 - ▶ MPI_Allreduce()
 - ▶ Scales as $\text{Log}(\#\text{nodes})$
- ▶ MG (-2%)
 - ▶ Exchange Halo (neighbor)
- ▶ SPMV (-0%)
 - ▶ Overlapped w/Compute



REPRODUCIBILITY

- ▶ Residual Variance (reported in output file)
 - ▶ zero = deterministic order of floating point operations
- ▶ GPU Supercomputers bitwise reproducible up to full scale
 - ▶ except with network hardware-acceleration enabled on Cray XC30
- ▶ Parallel Dot Product
 - ▶ Local GPU routines bitwise reproducible
 - ▶ MPI_Allreduce()
 - ▶ reproducible with default MPI implementation
 - ▶ Non-reproducible with network offload (hardware atomics)

REPRODUCIBILITY

- ▶ CRAY XC30 MPI_Allreduce()

- ▶ Default → reproducible results but lower performance

- ▶ Min MPI_Allreduce time: 0.0296645
 - ▶ Max MPI_Allreduce time: 0.153267
 - ▶ Avg MPI_Allreduce time: 0.0916832

- ▶ MPICH_USE_DMAPP_COL=1

- ▶ Min DDOT MPI_Allreduce time: 0.0379143
 - ▶ Max DDOT MPI_Allreduce time: 0.0379143
 - ▶ Avg DDOT MPI_Allreduce time: 0.0379143

- ▶ Residuals:

- 4.25079640861055e-08
 - 4.25079640861032e-08
 - 4.25079640861079e-08
 - 4.25079640861054e-08

POWER CONSUMPTION

- ▶ Piz Daint (5208 K20X)

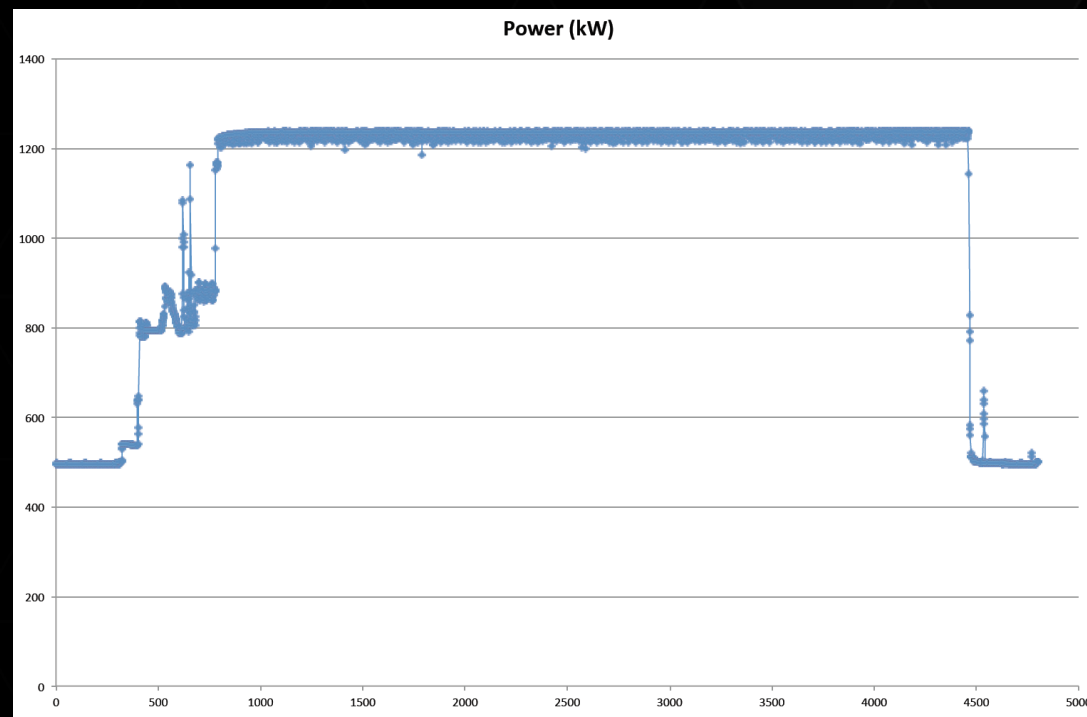
- ▶ 99 TF / 1232 kW

- ▶ 0.080 GF/W

- ▶ GK20A (Jetson TK1)

- ▶ 1.4 GF / 8.3 Watts

- ▶ 0.168 GF/W



PLATFORM COMPARISON

	MPI Tasks	# iteration	HPCG (GFlops)	Total Memory BW	HPCG per task	Ratio	Ratio RAW	HPCG - rank
Thiane-2A	46080	57	580109	14745600	12.59 GF	3.90%	4.40%	1
K	82944	51	426972	5308416	5.14 GF	8.00%	8.19%	2
Titan	18648	55	317216	4654540	17.01 GF	6.80%	7.48%	3
Piz-Daint	5208	55	97280	1299916	18.67 GF	7.40%	8.21%	5

Data from ISC14

CONCLUSIONS/ SUGGESTIONS

- ▶ (C) GPUs proven effective for HPL, especially for power efficiency
 - ▶ High flop rate
- ▶ (C) GPUs also very effective for HPCG
 - ▶ High memory bandwidth (Stacked memory will give a huge boost)
- ▶ (S) Reduce the required runtime from 1h to at least 100 iterations
- ▶ (S) Change metric: DOF/s?
- ▶ (S) Include yaml files in the list
- ▶ (S) Add power consumption?

ACKNOWLEDGMENTS

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