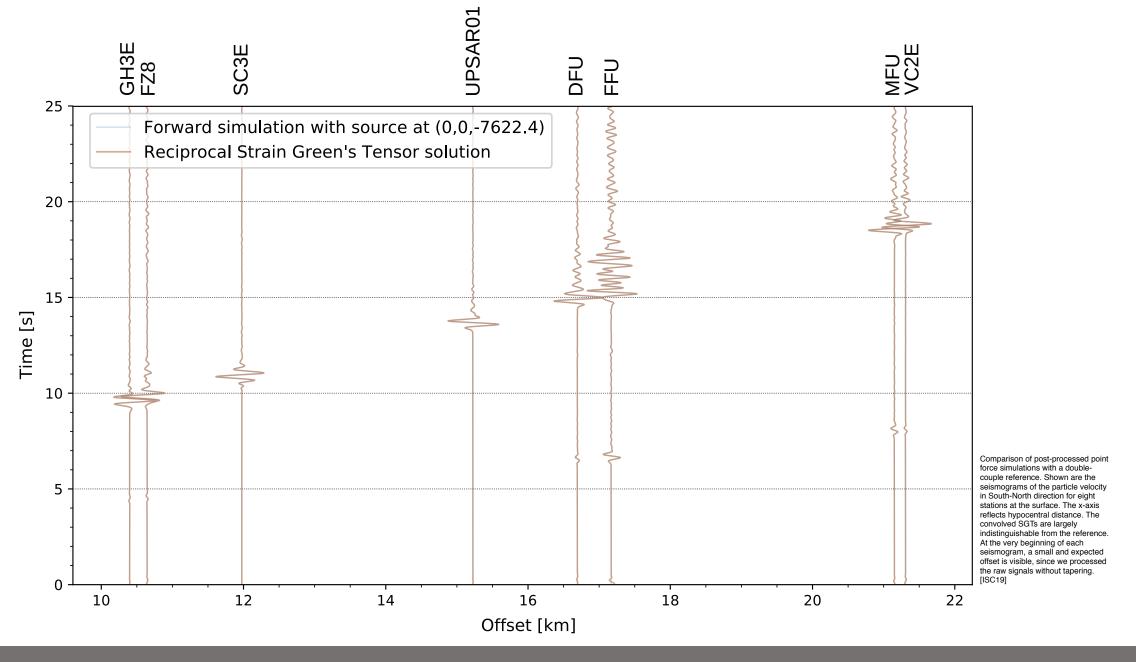


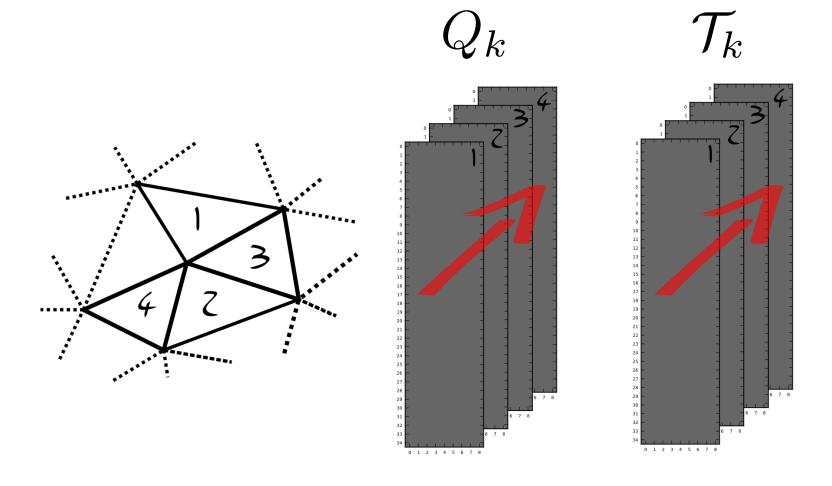
Time (s): 3.25



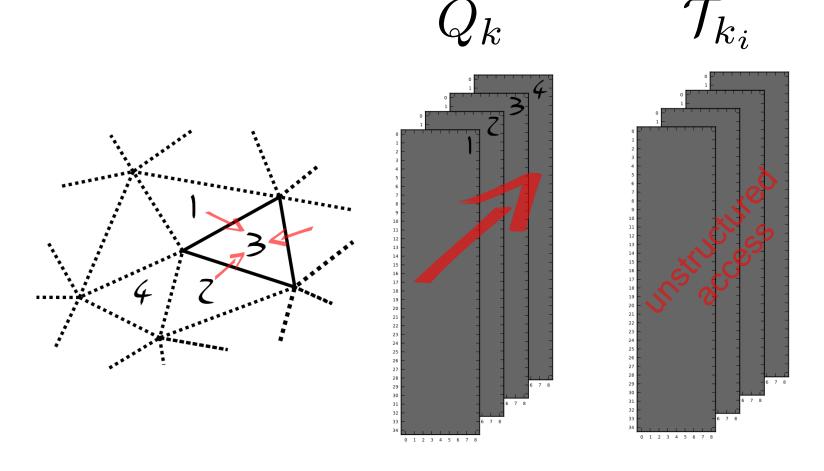


$$q_t + A^{x_1}q_{x_1} + A^{x_2}q_{x_2} + A^{x_3}q_{x_3} = 0$$

Local

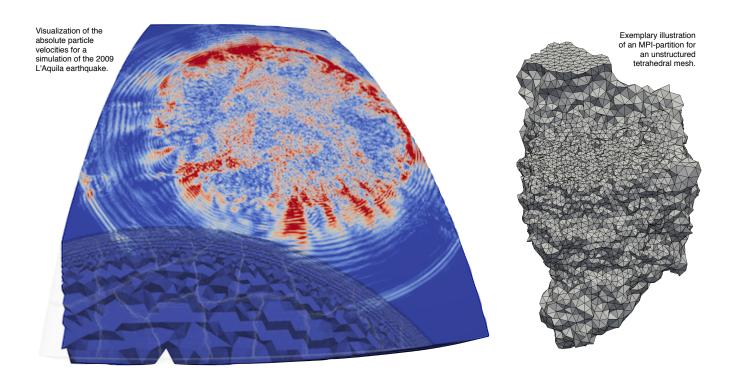


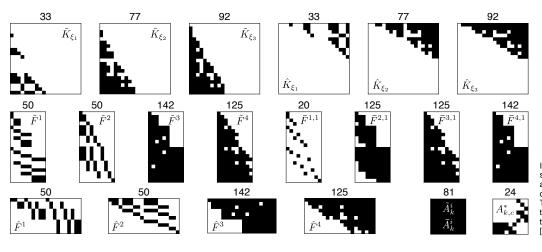
Neighboring



Solver

- Discontinuous Galerkin Finite Element Method (DG-FEM), ADER in time
- Full elastic wave equations in 3D and complex heterogeneous media
- Unstructured, conforming tetrahedral meshes
- Small sparse matrix operators in inner loops
- Compute bound (high orders)





Ilustration of all involves parse matrix patterns for thouth order ADER-DG discretization in EDGE. The numbers on top give the non-zero entries in the sparse matrices. Parco18]



Weak Scaling Runs

Year	System	Architecture	Nodes	Cores	Order	Precision	HW-PFLOPS	NZ-PFLOPS	NZ-%Peak
2014	SuperMUC	SNB	9216	147456	6	FP64	1.6	0.9	26.6
2014	Stampede	SNB+KNC	6144	473088	6	FP64	2.3	1.0	11.8
2014	Tianhe 2	IVB+KNC	8192	1597440	6	FP64	8.6	3.8	13.5
2015	SuperMUC 2	HSW	3072	86016	6	FP64	2.0	1.0	27.6
2016	Theta	KNL	3072	196608	4	FP64	1.8	1.8	21.5
2016	Cori 2	KNL	9000	612000	4	FP64	5.0	5.0	18.1
2018	AWS EC2	SKX	768	27648	5	FP32	1.1	1.1	21.2

A collection of weak scaling runs for elastic wave propagation with ADER-DG. The runs had similar but not identical configurations. Details are available from the given sources.

Explanation of the columns:

- · System: Name of the system or cloud service (last row).
- · Code-name of the used microarchitecture: Sandy Bridge (SNB), Ivy Bridge (IVB), Knights Corner (KNC), Haswell (HSW), Knights Landing (KNL), Skylake (SKX).
- · Nodes: Used number of nodes in the run.
- · Cores: Used number of cores in the run; includes host and accelerators cores for the heterogeneous runs.

- · Order: Used order of convergence in the ADER-DG solver. · Precision: Used floating point precision in the ADER-DG
- HW-PFLOPS: Sustained Peta Floating-Point Operations Per Stampede, Tianhe-2: [SC14] Second (PFLOPS) in hardware.
- NZ-PFLOPS: Sustained Peta Floating-Point Operations Per Theta, Cori: [ISC17] Second (PFLOPS) if only non-zero operations are counted, •AWS EC2: [ISC19] i.e., ignoring artificial operations, introduced through dense matrix operators on sparse matrices.
- · NZ-%Peak: Relative peak utilization, when comparing the machines' theoretical floating point performance to the sustained NZ-PFLOPS.

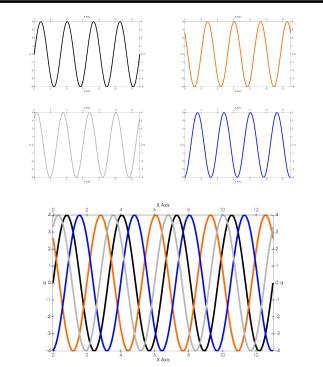
- · SuperMUC: [ISC14], [SC14]
- SuperMUC 2: [IPDPS16]





Introduction of "Mini-batches" for PDEs

Year	System		
2014	SuperMUC		
2014	Stampede		
2014	Tianhe 2		
2015	SuperMUC 2		
2016	Theta		
2016	Cori 2		
2018	AWS EC2		



Order	Precision	HW-PFLOPS	NZ-PFLOPS	NZ-%Peak
6	FP64	1.6	0.9	26.6
6	FP64	2.3	1.0	11.8
6	FP64	8.6	3.8	13.5
6	FP64	2.0	1.0	27.6
4	FP64	1.8	1.8	21.5
4	FP64	5.0	5.0	18.1
5	FP32	1.1	1.1	21.2

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- NZ-%Peak: Relative peak utilization, when comparing the machines' theoretical floating point performance to the sustained NZ-PFLOPS.

ouices.

- SuperMUC: [ISC14], [SC14]
- · Stampede, Tianhe-2: [SC14]
- SuperMUC 2: [IPDPS16]
- Theta, Cori: [ISC17]
- AWS EC2: [ISC19]



Cloud Computing

Micro-Benchmarks

Machine Setup

Performance Evaluation



Key Performance Indicators (KPIs)

KPI	KPI c5.18xlarge		m5.24xlarge	on-premises	
CSP	Amazon	Amazon	Amazon	N/A	
CPU name	8124M*	8124M*	8175M*	8180	
#vCPU (incl. SMT)	2x36	2x36	2x48	2x56	
#physical cores	2x18**	2x18**	2x24**	2x28	
AVX512 Frequency	≤3.0GHz	≤3.0GHz	≤2.5GHz	2.3GHz	
DRAM [GB]	144	192	384	192	
#DIMMs	2x10?	2x12?	2x12/24?	2x12	
spot \$/h	0.7	0.7	0.96	N/A	
on-demand \$/h	3.1	3.9	4.6	N/A	
interconnect [Gbps]	25***(eth)	25***/100***(eth)	25***(eth)	100(OPA)	

Publicly available KPIs for various cloud instance types of interest to our workload. Pricing is for US

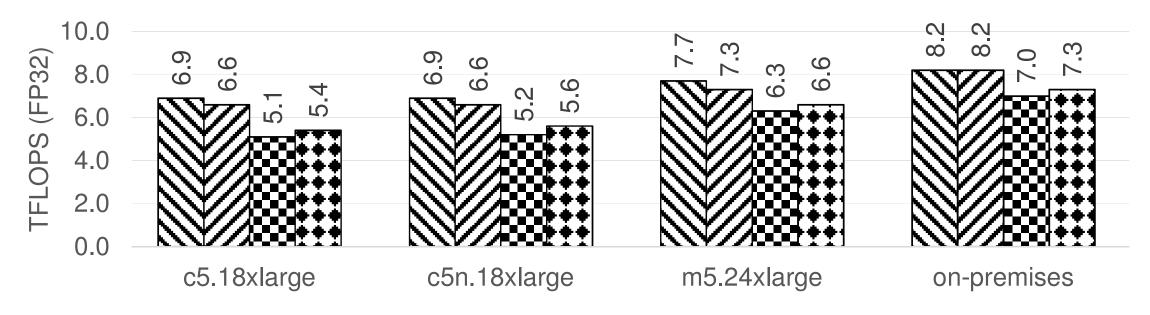
East at non-discount hours on Monday mornings (obtained on 3/25/19).

100Gbps for c5n.18xlarge reflects a recent update of the instance types (mid 2019).

*AWS CPU core name strings were retrieved using the "Isopu" command; "AWS physical cores are assumed from AWS's documentation, indicating that all cores are available to the user due to the Nitro Hypervisor; ***Supported in multi-flow scenarios (means multiple communicating processes per host).



Micro-Benchmarking: 32-bit Floating Point

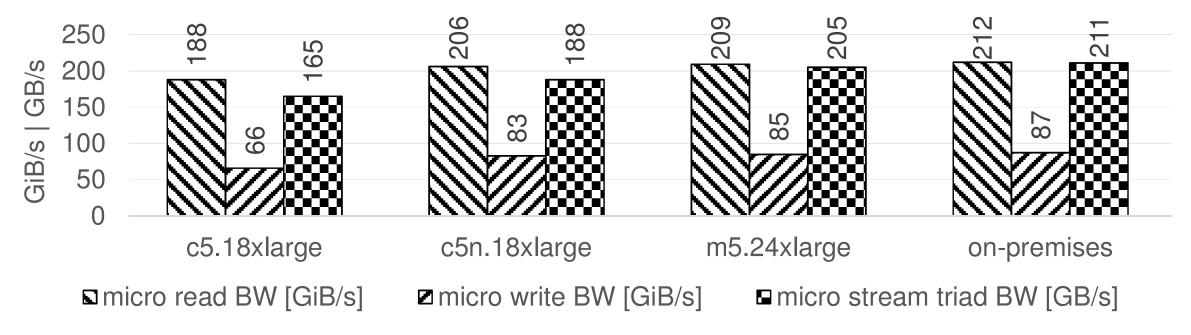


Sustained FP32-TFLOPS of various instance types: a) simple FMA instruction from register (micro FP32 FMA), b) an MKL-SGEMM call, spanning both sockets (SGEMM 2s), and c) two MKL-SGEMM calls, one per socket (SGEMM 1s). All numbers are compared to the expected AVX512 turbo performance (Paper PEAK).

on-premises: dual-socket Intel Xeon Platinum 8180, 2x12 DIMMs. [ISC19]

- □ Paper PEAK [TFLOPS]
- □ SGEMM 2s [TFLOPS]
- ☐ micro FP32 FMA [TFLOPS]
- 2x SGEMM 1s [TFLOPS]

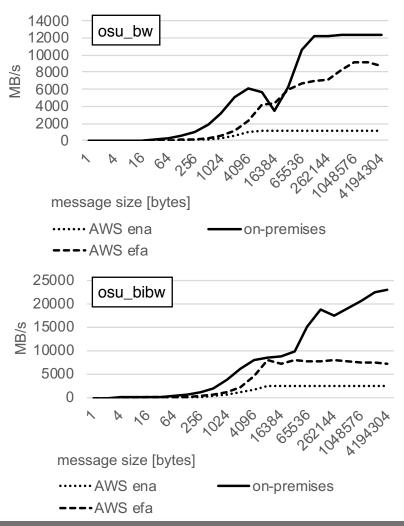
Micro-Benchmarking: Memory

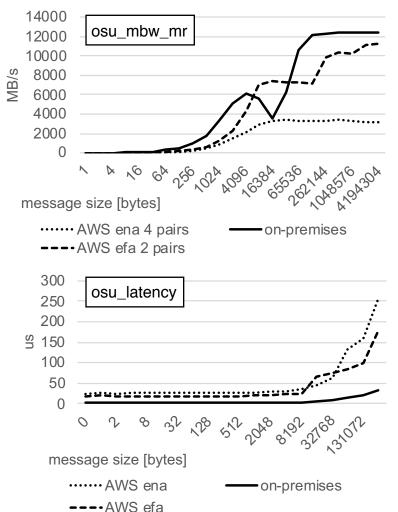


Sustained bandwidth of various instance types: a) a pure read-bandwidth benchmark (read BW), b) a pure write-bandwidth benchmark (write BW), and c) the classic STREAM triad with 2:1 read-to-write mix (stream triad BW), on-premises: dual-socket Intel Xeon Platinum 8180, 2x12 DIMMS, IISC191



Micro-Benchmarking: Network



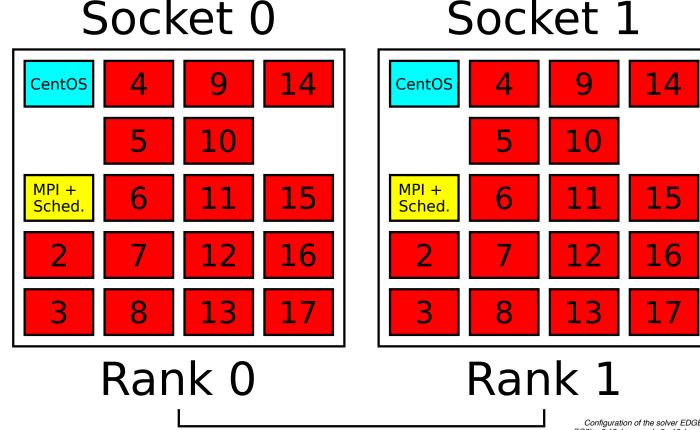


Interconnect performance of c5.18xlarge (AWS ena), c5n.18xlarge (AWS efa) and the on-premises, bare-metal system. Shown are results for the benchmarks osu_bw, osu_mbw_mr, osu_bibw and osu_latency (version 5.5). on-premises: dual-socket Intel Xeon Platinum 8180, 2x12 DIMMS, Intel OPA (100Gbps).



Machine Setup

- 1. Select instance type
- 2. Create machine image:
 - OS customization: core specialization, C-states, huge pages, TCP tuning, ...
 - System-wide installation of tools and dependencies
- 3. Create Slurm-based cluster:
 - Compute nodes / instances boot customized machine image
- 4. Run jobs as on every other supercomputer



2x 10Gbps flows

Configuration of the solver EDGE for AWS EC2's c5.18xlarge and c5n.18xlarge instance types. The first core of both sockets is reserved for the operating system. We spawn one MPI-rank per-socket for two flows per instances. The second core of every socket is reserved for our scheduling and MPI-progression thread. The remaining 16 cores of every socket are occupied by the 16 worker threads per rank.

Machine Setup

1. Select instance type

Center

2. Create machine image:

 OS customization: core specialization, C-states, huge pages, TCP tuning, ..

Vendor or Center

System-wide installation of tools and dependencies

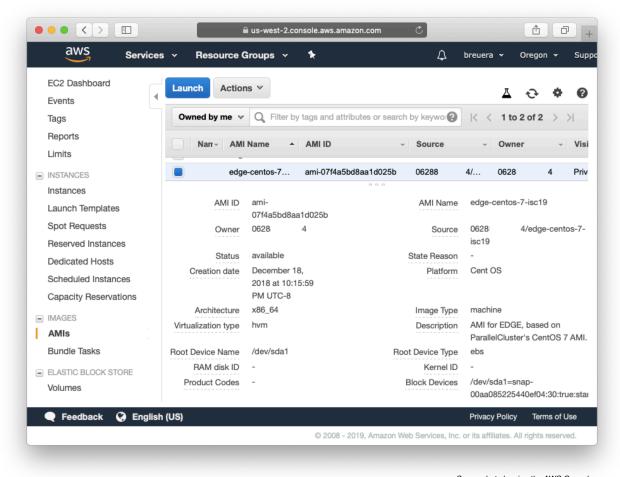
3. Create Slurm-based cluster:

 Compute nodes / instances boot customized machine image

Center

4. Run jobs as on every other supercomputer

User



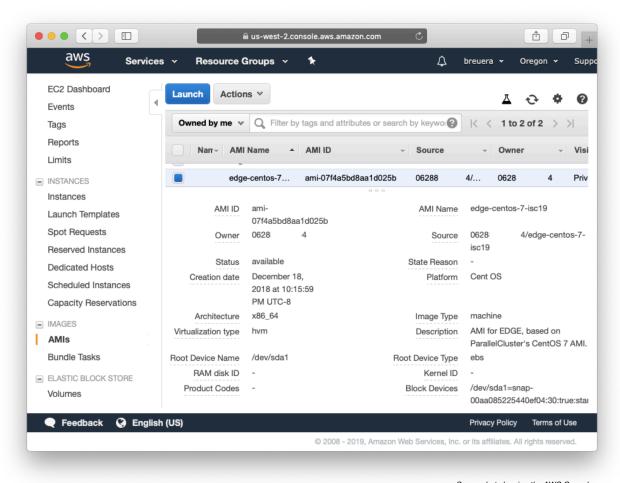
Screenshot showing the AWS Console for the Amazon Machine Image, used in [ISC19]'s large-scale simulations.



Machine Setup

- 1. Select instance type
- 2. Create machine image:
 - OS customization: core specialization, C-states, huge pages, TCP tuning, ..
 - System-wide installation of tools and dependencies
- 3. Create Slurm-based cluster:
 - Compute nodes / instances boot customized machine image
- 4. Run jobs as on every other supercomputer

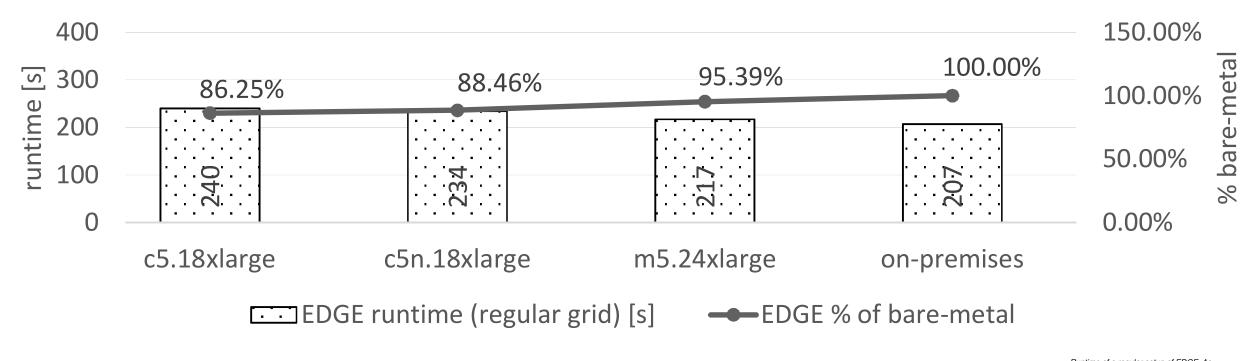
100% Open Source



Screenshot showing the AWS Console for the Amazon Machine Image, used in [ISC19]'s large-scale simulations.



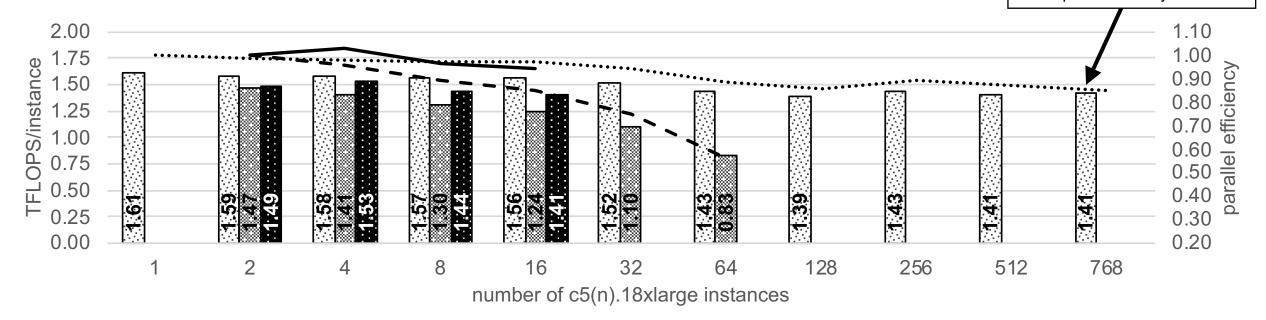
Cloud Virtualization vs. Bare Metal



Runtime of a regular setup of EDGE. As expected, all cloud instances are slower than the top-bin bare-metal machine. AWS instances are within 85% of the on-premises performance. on-premises: dual-socket Intel Xeon Platinum 8180, 2x12 DIMMS, Intel OPA (100Gbps).

Petascale Cloud Computing

1.09 non-zero FP32-PFLOPS 21.2% peak efficiency @2.9GHz



weak-scaling TFLOPS/instance (c5.18xlarge) strong-scaling TFLOPS/instance (c5n.18xlarge)

strong-scaling efficiency (c5.18xlarge)

strong-scaling TFLOPS/instance (c5.18xlarge)

······ weak-scaling efficiency (c5.18xlarge)

strong-scaling efficiency (c5n.18xlarge)

Weak and strong scalability of EDGE in AWS EC2 on c5.18xlarge and c5n. 18xlarge instances. We sustained 1.09 PFLOPS in weak-scaling on 768 c5.18xlarge instances. This elastic high performance cluster contained 27,648 Skylake-SP cores with a peak performance of 5 PFLOPS.



Part of a Comprehensive Approach

- Machine:
 - Hardware selection
 - OS customization
 - HPC Environment



- Single Node:
 - Kernels
 - Custom OpenMP and load balancing
 - Memory Layout
- Multi Node:
 - Overlapping communication and computation
 - Prioritization of crucial work
 - Communication "as is", no additional MPIbuffers

- Algorithmic: Clustered Local Time Stepping (LTS), fused simulations
- Software Engineering:
 - CI/CD, continuous verification
 - Workflow automation
 - Software and data sharing
- Modeling and Simulation:
 - Model extensions
 - Surface meshing
 - Volume meshing
 - Mesh annotations
- Data Analysis:
 - Verification
 - SGT assembly and processing



Outlook: Beyond Petascale

Year	System	Architecture	Nodes	Cores	Order	Precision	HW-PFLOPS	NZ-PFLOPS	NZ-%Peak
2018	AWS EC2	SKX	768	27648	5	FP32	1.1	1.1	21.2
2016	Cori 2	KNL	9000	612000	4	FP64	5.0	5.0	18.1
2015	SuperMUC 2	HSW	Gurrent:	2010			Outlook:	10	07.0
2016	Theta		25Gbps c5.18xlarge (limited to 20Gbps in our configuration) - 100Gbps network closes gap to on-premises solutions						
2014		IVB+KN(2							
2014	Stampede		• Spot-instances and us-west-2 Cloud is (much) bigger than our						
2014	SuperMUC	SNB (Oregon)				run (general is the limit?	purpose CPU	s); what

- es gap to
- than our CPUs); what

A collection of weak scaling with ADER-DG. The runs had similar but not identica configurations. Details are available from the given sources.

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- · Stampede, Tianhe-2: [SC14] • SuperMUC 2: [IPDPS16]
- Theta, Cori: [ISC17]
- AWS EC2: [ISC19]



References

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This work heavily used contributions of many authors to open-source software.

This software includes, but is not limited to: ASan (https://clang.llvm.org/docs/AddressSanitizer.html, debugging), AWS Parallel Cluster (https://github.com/aws/aws-parallelcluster, clusters in AWS), Catch (https://github.com/philsquared/Catch, unit tests), CentOS (https://www.centos.org, cloud OS), CGAL (http://www.cgal.org, surface meshes), Clang (https://clang.llvm.org/, compilation), Cppcheck (http://cppcheck.sourceforge.net/, static code analysis), Easylogging+ (https://github.com/easylogging/, logging), ExprTk (https://partow.net/programming/exprtk, expression parsing), GCC (https://gcc.gnu.org/, compilation), Git (https://git-scm.com, versioning), Git LFS (https://git-lfs.github.com, versioning), Gmsh (http://gmsh.info/, volume meshing), GoCD (https://www.gocd.io/, continuous delivery), HDF5 (https://www.hdfgroup.org/HDF5/, I/O), jekyll (https://jekyllrb.com, homepage), LIBXSMM (https://github.com/hfp/libxsmm, matrix kernels), METIS (http://glaros.dtc.umn.edu/gkhome/metis/overview, partitioning), MOAB (http://sigma.mcs.anl.gov/moab-library/, mesh interface), NetCDF (https://www.unidata.ucar.edu/software/netcdf/, I/O), ObsPy (https://github.com/obspy/obspy/wiki, signal analysis), OpenMPI (https://www.openmpi.org, cloud MPI), ParaView (http://www.paraview.org/, visualization), pugixml (http://pugixml.org/, XML interface), Read the Docs (https://www.openmpi.org, Container virtualization), surm-GCP (https://github.com/schedMD/slurm-gcp, clusters in GCP), TF-MISFIT GOF CRITERIA (http://www.nuquake.eu, signal analysis), UCVMC (https://github.com/SCECcode/UCVMC, velocity model), Valgrind (http://valgrind.org/, memory debugging), Visit (https://wci.llnl.gov/simulation/computer-codes/visit, visualization).



ISC17

- Theta (ALCF @ ANL)
 - Cray XC40 (early science)
 - 3,200 Intel Xeon Phi 7230 at 1.3 GHz (with Intel Turbo Boost enabled)
 - 16 GB of in-package HBM and 192GB of DDR4 RAM
- Cori Phase II (NERSC @ LBNL)
 - Cray XC40
 - 9,304 Intel Xeon Phi 7250 68-core processors at 1.4 GHz (with Intel Turbo Boost enabled)
 - 16 GB of in-package HBM and 96 GB of DDR4 RAM



$$\mathcal{T}_{k}(t_{0}, \hat{t}, \Delta t) = \int_{\hat{t}}^{\hat{t} + \Delta t} Q_{k}(t) dt = \sum_{d=0}^{\mathcal{O}-1} \frac{(\hat{t} + \Delta t)^{d+1} - \hat{t}^{d+1}}{(d+1)!} \cdot \frac{\partial^{d}}{\partial t^{d}} Q_{k}(t_{0})$$

Volume

$$\mathcal{V}_k(\mathcal{T}_k) = \sum_{c=1}^3 A_k^{\xi_c} \mathcal{T}_k K^{\xi_c} M^{-1}$$

$$\mathcal{F}_{k,i}^{-}(\mathcal{T}_k) = \frac{|S_i|}{|J_k|} \hat{A}_{k,i}^{-} \mathcal{T}_k F^{-,i} M^{-1} \qquad \mathcal{F}_{k,i}^{+}(\mathcal{T}_{k_i}) = \frac{|S_i|}{|J_k|} \hat{A}_{k,i}^{+} \mathcal{T}_{k_i} F^{+,i,j,k} M^{-1}$$

Local Update

$$Q_k^{*,n_k+1} = Q_k^{n_k} + \mathcal{V}_k(\mathcal{T}_k) - \sum_{i=1}^4 F_{k,i}^-(\mathcal{T}_k)$$
$$Q_k^{n_k+1} = Q_k^{*,n_k+1} - \sum_{i=1}^4 \mathcal{F}_{k,i}^+(\mathcal{T}_{k_i})$$

Neighboring Update

$$Q_k^{n_k+1} = Q_k^{*,n_k+1} - \sum_{i=1}^4 \mathcal{F}_{k,i}^+ (\mathcal{T}_{k_i})$$

Time
$$\mathcal{T}_k(t_0,\hat{t},\Delta t) = \int_{\hat{t}}^{\hat{t}+\Delta t} Q_k(t) \, \mathrm{d}t = \sum_{d=0}^{\mathcal{O}-1} \frac{\left(\hat{t}+\Delta t\right)^{d+1} - \hat{t}^{d+1}}{(d+1)!} \cdot \frac{\partial^d}{\partial t^d} Q_k(t_0)$$
 Volume
$$\mathcal{V}_k(\mathcal{T}_k) = \sum_{1}^{3} A_k^{\xi_i} \mathcal{T}_k K^{\xi_c} M^{-1}$$
 Flux
$$\mathcal{F}_{k,i}^-(\mathcal{T}_k) = \frac{|S_i|}{|J_k|} \hat{A}_k^-, \mathcal{T}_k \mathcal{F}^{-,i} M^{-1} \qquad \mathcal{F}_{k,i}^+(\mathcal{T}_{k_i}) = \frac{|S_i|}{|J_k|} \hat{A}_k^+, \mathcal{T}_{k_i} \mathcal{F}^{+,i,j,k} M^{-1}$$

Neighboring Update

$$Q_k^{*,n_k+1} = Q_k^{n_k} + \mathcal{V}_k(\mathcal{T}_k) - \sum_{i=1}^4 F_{k,i}^-(\mathcal{T}_k)$$
$$Q_k^{n_k+1} = Q_k^{*,n_k+1} - \sum_{i=1}^4 \mathcal{F}_{k,i}^+(\mathcal{T}_{k_i})$$