Many-Cores for the Masses: A Year With the Cori System at NERSC





Richard Gerber Jack Deslippe Intel HPC Developer Conference November, 2017





# **NERSC: Mission HPC for DOE Office of Science**





Office of Science

Largest funder of physical science research in U.S.



Bio Energy, Environment



Particle Physics, Astrophysics

Office of Science



Computing



**Nuclear Physics** 



Materials, Chemistry, Geophysics



Fusion Energy, Plasma Physics

6,000 users, 700 projects, 700 codes, 48 states, 40 countries, universities & national labs



# Focus on Science

017 to date

NERSC

NERSC users produce publish more than any other center in the world\*; ~2K/year

1,036 citations via Web of Science in 2017 so far (underestimate!)





5 in Nature 30 in Nature Comm. 70 in 12 journals





\* as far as I can tell



# **High Performance Computing Systems**



#### Cori

9,300 Intel Xeon Phi "KNL" manycore nodes 2,000 Intel Xeon "Haswell" nodes 700,000 processor cores, 1.2 PB memory Cray XC40 / Aries Dragonfly interconnect 30 PB Lustre Cray Sonexion scratch FS 1.5 PB Burst Buffer

Haswell: ~1 B NHrs/yr; KNL: ~6 B NHrs/yr



#6 on June 2017 Top 500 list



#### Edison

5,560 Ivy Bridge Nodes / 24 cores/node 133 K cores, 64 GB memory/node Cray XC30 / Aries Dragonfly interconnect 6 PB Lustre Cray Sonexion scratch FS Edison: ~2 B NHrs/yr





# **Change Has Arrived**



Driven by power consumption and heat dissipation toward lightweight cores







KNL: 215-230 W 2-socket Haswell: 270 W

Cori, a 30 PFlop system, will be a boon to science in the U.S. because of new capabilities, but the Intel Xeon Phi many-core architecture will require a code modernization effort to use efficiently.







Adoption of KNL has been good and Cori KNL nodes are fully used by DOE Office of Science Researchers

- 150 projects have used > 1 M NERSC Hours
- 233 projects have used > 100 K NERSC Hrs
- Still leaves ~500 projects to move over

Office of Science

- 32% of hours use > 1,024 nodes (69K cores)
- Gordon Bell submissions using Cori KNL
  Massively Parallel 3D Image Reconstruction, Wang et al.



Cori KNL Hours Used Jan-Aug 2017



# **High Impact Science at Scale Projects**





Optical Properties of Materials, Louie, UC Berkeley Strangeness and Electric Charge Fluctuations in Strongly Interacting Matter, Karsch, Brookhaven





M8 Earthquake on the San Andreas Fault, Goulet, USC Earthquake Center



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Asymmetric Effects in Plasma Accelerators, Vay, Berkeley Lab



# **Deep Learning on Cori KNL**



### **NERSC** is actively exploring Deep Learning for Science

- Collaborating with leading vendors to optimize and deploy stack
- Collaborating with leading research institutions to develop methods
- Drive real science use cases

### Deep Learning at 15 PF on NERSC Cori (Cray + Intel KNL)

- Trained in 10s of minutes on 10 terabyte datasets, millions of Images
- 9600 nodes, optimized on KNL with IntelCaffe and MKL (NERSC / Intel collaboration)
- Synch + Asynch parameter update strategy for multi-node scaling (NERSC / Stanford)



See Prabhat's plenary talk Sunday morning

Identified extreme climate events using supervised (left) and semisupervised (right) deep learning. Green = ground truth, Red = predictions (confidence > 0.8). [NIPS 2017]







# How Did We Get Here?





# NESAP - NERSC Exascale Science Apps Program











# What is different about Cori?



### Edison ("Ivy Bridge):

- 5576 nodes
- 24 physical cores per node
- 48 virtual cores per node
- 2.4 3.2 GHz
- 8 double precision ops/cycle
- 64 GB of DDR3 memory (2.5 GB per physical core)
- ~100 GB/s Memory Bandwidth

## Cori ("Knights Landing"):

- 9304 nodes
- 68 physical cores per node
- 272 virtual cores per node
- 1.4 1.6 GHz
- 32 double precision ops/cycle
- 16 GB of fast memory 96GB of DDR4 memory
- Fast memory has 400 500 GB/s
- No L3 Cache







- Goal: Prepare DOE Office of Science users for many core
- Partner closely with ~20 application teams and apply lessons learned to broad NERSC user community.
- 20 applications cover (or serve as proxies for) > 50% of NERSC hours used in 2016

Activities:

Science





Energy-Efficient Processors Have Multiple Hardware Features to Optimize Against:

- Many (Heterogeneous) Cores
- Bigger Vectors
- New ISA
- Multiple Memory Tiers

#### It is easy for users to get bogged down in the weeds:

- How do you know what KNL hardware feature to target?
- How do you know how your code performs in an absolute sense and when to stop?





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NERSO







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# NERSC has developed tools and strategy for users to answer these questions:

- Designed simple tests that demonstrate code limits
- Use roofline as an optimization guide
- Training and documentation hub targeting all users





### **Tools CoDesign**





Intel Vector-Advisor Co-Design - Collaboration between NERSC, LBNL Computational Research, Intel





# **Example: WARP (Accelerator Modeling)**







#### **Optimizations:**

- 1. Add tiling over grid targetting L2 cache on both Xeon + Xeon-Phi Systems
- 2. Apply particle sorting + vectorization over particles (requires a number of datastructure changes)







KNL Roofline

# **Example: WARP (Accelerator Modeling)**







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# **KNL** Performance











# **Preliminary NESAP Code Performance on KNL**







### **NESAP MCDRAM Effects**









### **NESAP VPU Effects**



AVX512 vs AVX2







# What did we learn?



- It is crucial to understand what limits performance for your code/kernels. Tools like Advisor are necessary.
- To get good performance on KNL. One typically needs good task/thread scaling and depending on algorithm:
  - a) efficient vectorization (Codes with high AI)
  - b) efficient use of the MCDRAM (Codes with low AI)
  - c) both (Codes with AI near 1)
- The lack of an L3 cache on KNL can make cache blocking for L1/L2 more important. Particularly in latency-sensitive apps (e.g. indirect indexing)
- Cache mode provides nearly the same performance as flat mode (with directives) for most applications. However, cache-conflicts can be an issue with some apps.
- MPI apps tend to stop scaling at the same number of ranks on Xeon and Xeon-Phi (often characterized by the algorithm). This translates to lower node counts on Xeon-Phi. Additional, parallelism needs to be exploited usually expressed as OpenMP.





# First anisotropic, 3-pt correlation computation on 2B Galaxies from Outer Rim Simulation

 Solves an open problem in cosmology for the next decade (LSST will observe 10B galaxies)

3-Pt Correlation On 2B Galaxies Recently Completed

NESAP For Data Prototype (Galactos)

- Can address questions about the nature of darkenergy and gravity
- Novel O(N<sup>2</sup>) algorithm based on spherical harmonics for 3-pt correlation

#### Scale:

on Cori

- 9600+ KNL Nodes (Significant Fraction of Peak)





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# END, Thank you!









