Introduction to HPC2N, Kebnekaise and High Performance Computing

Mirko Myllykoski mirkom@cs.umu.se and others at HPC2N

Department of Computing Science and HPC2N Umeå University

22. January 2020





HPC2N





High Performance Computing Center North (HPC2N) is a national center for Scientific and Parallel Computing



Provides state-of-the-art resources and expertise:





- Provides state-of-the-art resources and expertise:
 - Scalable and parallel HPC





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 - Large-scale storage facilities (PFS-Lustre, SweStore, Tape)





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 - Software and advanced support
 - International network for research and development



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- Five partners:
 - Luleå University of Technology
 - Mid Sweden University
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 - Swedish University of Agricultural Sciences SLU
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- Involved in several projects and collaborations
 - EGI, PRACE, EISCAT, eSSENCE, NOSEG, SNIC Science Cloud, NLAFET, ...



User support (primary, advanced, tailored)

- Research group meetings @ UmU
- Also the partner sites



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User training and education program

- 0.5 3 days; ready-to-run exercises
- Introduction to HPC2N and Kebnekaise
- Parallel programming and tools (e.g., OpenMP, MPI, debugging, performance analyzers, Matlab, R, MD simulation, Deep Learning, GPU, ...)



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- NGSSC / SeSE & university courses
- Workshops and seminars



Management

- Bo Kågström, director
- Lena Hellman, administrator
- Björn Torkelsson, technical coordinator



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System and support

- Erik Andersson
- Birgitte Brydsö
- Niklas Edmundsson (Tape coord)
- Ingemar Fällman
- Magnus Jonsson
- Roger Oscarsson
- Åke Sandgren
- Mattias Wadenstein (NeIC, Tier1)
- Lars Viklund





HPC2N provides advanced level support (tier-3 and tier-4) in the form of application experts:

Jerry Eriksson Profiling, Machine learning (DNN), MPI, OpenMP, OpenACC



Jerry Eriksson	Profiling,	Machine	learning	(DNN),	MPI,
	OpenMP, OpenACC				
Mirko Myllykoski	High performance computing, numerical linear algebra, CUDA, OpenCL, task parallelism				



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 Contact through support at HPC2N (create a ticket, tier-3 and tier-4 tickets are forwarded to suitable application experts)



HPC2N (users by discipline)

Users from several scientific disciplines:

- Biosciences and medicine
- Chemistry
- Computing science
- Engineering
- Materials science
- Mathematics and statistics
- Physics including space physics
- Deep Learning and AI in general (several new projects)



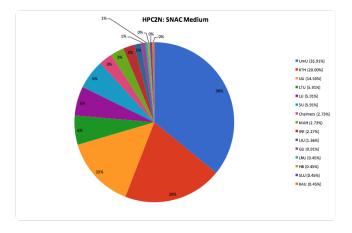
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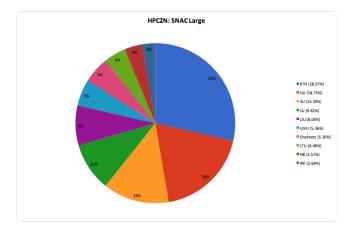
HPC2N (medium users by university)



Projects with allocations at HPC2N: 2014-01-01 to 2016-05-30



HPC2N (large users by university)



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Kebnekaise is the latest supercomputer at HPC2N

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- Kebnekaise was
 - delivered by Lenovo and
 - installed during the summer 2016
- Kebnekaise was opened up for general availability on November 7, 2016
- In 2018, Kebnekaise was extended with
 - ▶ 52 Intel Xeon Gold 6132 (Skylake) nodes, as well as
 - 10 NVidian V100 (Volta) GPU nodes



Kebnekaise (compute nodes)

	Name	#	Description
-	Compute	130	Intel Xeon E5-2690v4, 2 x 14 cores,
Compute	432	128 GB, FDR Infiniband	



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KNL	36	Intel Xeon Phi 7250 (Knight's Landing),
		68 cores, 192 GB, 16 GB MCDRAM,
		FDR Infiniband



Kebnekaise (GPU nodes)

Name	#	Description
2xGPU	32	Intel Xeon E5-2690v4, 2 x 14 cores, 128 GB, FDR Infiniband, 2 x NVidia K80 4 x 2496 CUDA cores, 4 x 12 GB VRAM



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2xGPU		Intel Xeon E5-2690v4, 2 x 14 cores,
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4xGPU		Intel Xeon E5-2690v4, 2 x 14 cores,
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		128 GB, FDR Infiniband,
		4 x NVidia K80
		8 x 2496 CUDA cores, 8 x 12 GB VRAM
GPU-volta	10	Intel Xeon Gold 6132, 2 x 14 cores,
		192 GB, EDR Infiniband,
		2 x NVidia V100,
		2×5120 CUDA cores, 2×16 GB VRAM,
		2 x 640 Tensor cores



602 nodes in 15 racks



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▶ 19288 cores (of which 2448 cores are KNL-cores)

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- 71 switches (Infiniband, Access and Managment networks)



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- 629 TFlops/s Linpack (all parts, except expansion)
 - 86% of Peak performance





Basically five types of storage are available at HPC2N:

Home directory

Only accessible from the login nodes, backed up regularly



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 - Only accessible from the **login nodes**, backed up regularly
- Center Storage Parallel file system (fast discs)
 - Accessible from login and compute nodes



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- Per node local scratch space
 - about 170GB, SSD, only during the lifetime of the batch job



High Performance Computing (definition)

"High Performance Computing most generally refers to the practice of **aggregating computing power** in a way that delivers much **higher performance** than one could get out of a typical desktop computer or workstation in order to **solve large problems** in science, engineering, or business."¹

¹https://insidehpc.com/hpc-basic-training/what-is-hpc/



High Performance Computing (opening the definition)

Aggregating computing power

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- Compared to 4 cores in a modern laptop



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Solve large problems

- When does a problem become large enough for HPC?
- Are there other reasons for using HPC resources?



High Performance Computing (large problems)

A problem can be large for two main reasons:

- 1. Execution time: The time required to form a solution to the problem is very long
- 2. Memory / storage use: The solution of the problem requires a lot of memory and/or storage



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- The former can be remedied by increasing the performance
 - More cores, more nodes, GPUs, ...
- The latter by adding more memory / storage
 - More memory per node (including large memory nodes), more nodes, ...
 - Large storage solutions, ...



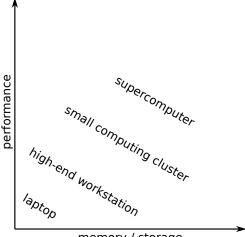
High Performance Computing (what counts as HPC)

performance

memory / storage



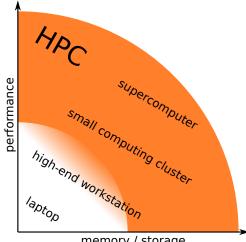
High Performance Computing (what counts as HPC)



memory / storage



High Performance Computing (what counts as HPC)



memory / storage





Specialized (expensive) hardware

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 - HPC2N holds licenses for several softwares
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- Support and documentation



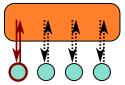
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 - Shared memory: Single memory space for all data.

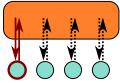


- Everyone can access the same data
- Straightforward to use



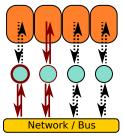
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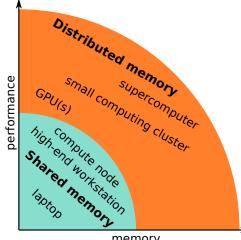
Distributed memory: Multiple distinct memory spaces.



- Everyone has direct access only to the local data
- Requires communication



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memory



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 - 3. Distributed memory: MPI,
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 - Work distribution, coordination (synchronization, etc),
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- GPUs: CUDA, OpenCL, OpenACC, OpenMP, ...
 - Many lightweight streams of operations



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 - Data distribution across memory spaces and movement



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 - Data distribution, storage, . . .
- ► GPUs: MAGMA, TensorFlow, ...
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End (questions?)

Questions?

