Introduction to HPC2N, Kebnekaise and HPC

Birgitte Brydsö, Pedro Ojeda May, and others at HPC2N

> HPC2N Umeå University

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







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 A part of National Academic Infrastructure for Supercomputing in Sweden (NAISS)









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

► Scalable and parallel **HPC**





- Scalable and parallel HPC
- Large-scale storage facilities (Project storage (Lustre), SweStore, Tape)





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 - Primary, advanced, dedicated
 - Application Experts (AEs)







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- Support
 - Primary, advanced, dedicated
 - Application Experts (AEs)
- International network for research and development







HPC2N (partners)

HPC2N has five **partners**:

- Luleå University of Technology
- ► Mid Sweden University
- Swedish Institute of Space Physics
- Swedish University of Agricultural Sciences (SLU)
- Umeå University







HPC2N (funding)

► Funded mainly by **Umeå University**, with contributions from the **other HPC2N partners**









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- Involved in several projects and collaborations
 - DDLS, EGI, EISCAT, eSSENCE, NOSEG, Swedish Science Cloud, . . .







- User support (primary, advanced, dedicated)
 - Research group meetings @ UmU
 - Also at 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

- Paolo Bientinesi, director
- Björn Torkelsson, deputy director
- Lena Hellman, administrator





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- Research Engineers under DDLS, HPC2N/SciLifeLab
 - System Developer, IT
 - Data Engineer
 - Data Steward







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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







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Contact through regular support

► If you have a specific problem/question and/or need consultation (up to 100 h)







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 artificial intelligence







HPC2N (users by discipline, largest users)

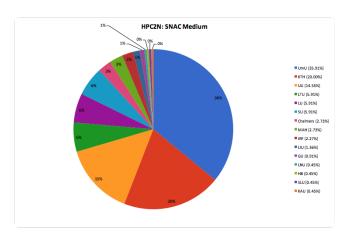
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HPC2N (medium users by university)



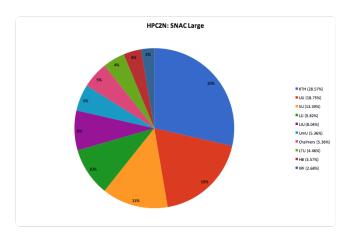
Projects with allocations at HPC2N: 2014-01-01 to 2016-05-30







HPC2N (large users by university)



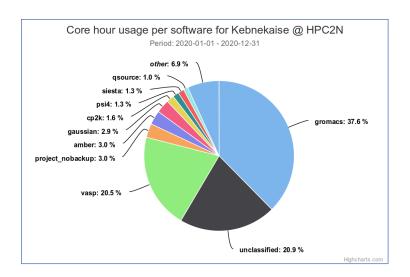
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HPC2N (users by software)









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- ▶ In 2023, Kebnekaise was **extended** with
 - 2 dual NVIDIA A100 GPU nodes
 - one many-core AMD Zen3 CPU node







Kebnekaise (compute nodes)

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







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Compute-skylake	32	192 GB, EDR Infiniband, AVX-512







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Compute-skylake		192 GB, EDR Infiniband, AVX-512
C . AMD 7 2		AMD Zen3 (EPYC 7762), 2 x 64 cores,
Compute-AMD Zen3	1	1 TB, EDR Infiniband







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Compute-skylake	52	Intel Xeon Gold 6132, 2×14 cores, 192 GB , EDR Infiniband, AVX-512
Compute-AMD Zen3	1	AMD Zen3 (EPYC 7762), 2 x 64 cores, 1 TB , EDR Infiniband
Large Memory	20	Intel Xeon E7-8860v4, 4 x 18 cores, 3072 GB, EDR Infiniband







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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4xGPU		Intel Xeon E5-2690v4, 2 x 14 cores,
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	4	4 x NVidia K80
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GPU-volta		Intel Xeon Gold 6132, 2 x 14 cores,
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GPU-volta	10	2 x NVidia V100,
		2 x 5120 CUDA cores, 2 x 16 GB VRAM,
		2 x 640 Tensor cores
2 × A100		AMD Zen3 (AMD EPYC 7413), 2 x 24 cores,
		512 GB, EDR Infiniband,
	2	2 x NVidia A100,
		2 x 6912 CUDA cores,
		2 x 432 Tensor cores







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







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 - ► Tape Storage
 - Backup
 - Long term storage







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- ▶ I will cover more details in the next section, where we go more into detail about HPC2N and Kebnekaise.





HPC

What is HPC?







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? (Memory, software, support, etc.)

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High Performance Computing (large problems)

- A problem can be large for two main reasons:
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 - ► More cores, more nodes, GPUs, ...





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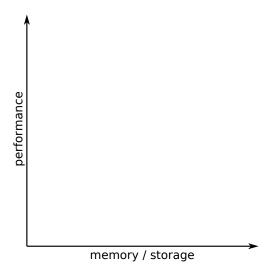
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- The latter by adding more memory / storage
 - More memory per node (including large memory nodes), more nodes, . . .
 - Kebnekaise: 128GB 192GB, 512GB, 3TB
 - Large storage solutions, . . .







High Performance Computing (what counts as HPC)

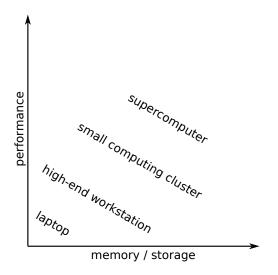








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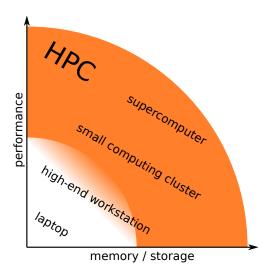








High Performance Computing (what counts as HPC)









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- Support and documentation







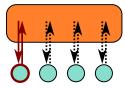
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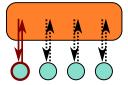


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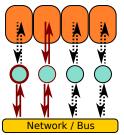




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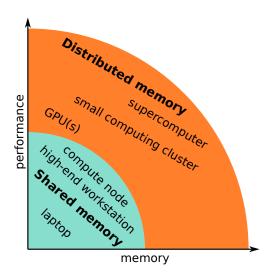
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- Distributed memory: Multiple distinct memory spaces.



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













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 - Work distribution, coordination (synchronization, etc), . . .
 - Data distribution across memory spaces and movement







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 - ▶ Data distribution, storage, . . .
- ► GPUs: MAGMA, TensorFlow, ...
 - ► Load correct toolchain etc
 - Allocate correct number of cores and GPUs, configure software to use correct number of cores and GPUs, . . .







End (questions?)

Questions?



