

#### ADVANCING GEOSCIENCE THROUGH LARGE-SCALE AI WITH SUPERCOMPUTING FOR EARTH OBSERVATION AND REMOTE SENSING

4<sup>th</sup> Summer School on High-Performance and Disruptive Computing in Remote Sensing I 4 June 2024 | Santiago de Compostela (Spain)

GABRIELE CAVALLARO (WWW.GABRIELE-CAVALLARO.COM) HEAD OF SIMULATION AND DATA LAB "AI AND ML FOR REMOTE SENSING", JÜLICH SUPERCOMPUTING CENTRE (FORSCHUNGSZENTRUM JÜLICH) ADJUNCT ASSOCIATE PROFESSOR, SCHOOL OF ENGINEERING AND NATURAL SCIENCES, UNIVERSITY OF ICELAND CO-CHAIR OF IEEE GRSS EARTH SCIENCE INFORMATICS TECHNICAL COMMITTEE (ESI TC)









#### OUTLINE

**1. Evolution of Computing** 

2. Machine Learning drives Computational Demands

**3.** Supercomputing and Parallel Programming

4. Jülich Supercomputing Centre

#### OUTLINE

#### **1. Evolution of Computing**

2. Machine Learning drives Computational Demands

3. Supercomputing and Parallel Programming

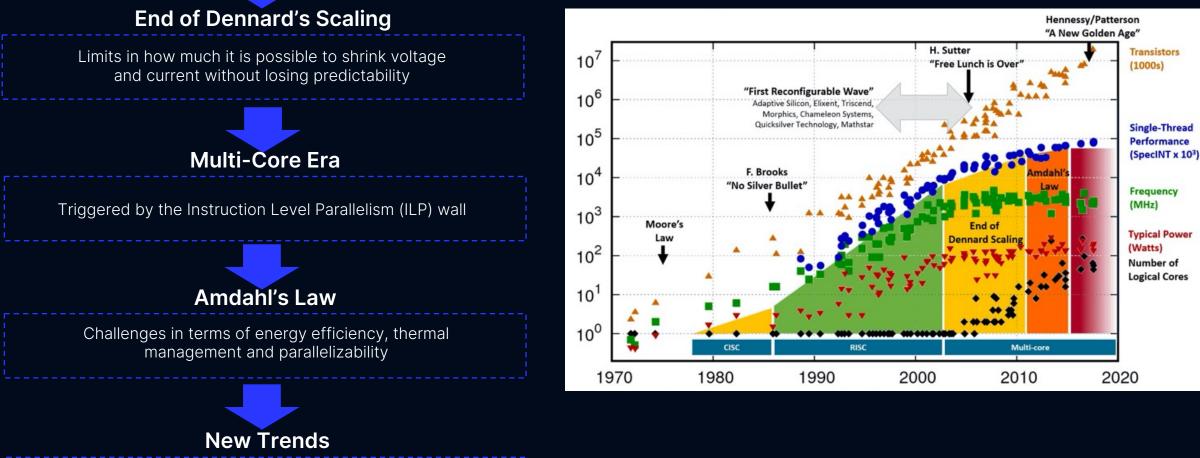
4. Jülich Supercomputing Centre

# COMPUTING TECHNOLOGIES HAVE EVOLVED IN RECENT DECADES

#### Moore's Law

Drove the semiconductor industry to cram more and more transistors and logic into the same volume

#### MICROPROCESSOR TREND DATA



To obtain the best performance-cost-energy tradeoffs for defined tasks

J. L. Hennessy, D. A. Patterson, "A New Golden Age for Computer Architecture", in Communications of the ACM, vol. 62 no. 2, pp. 48-60, 2019, https://doi.org/10.1145/3282307

Hennessy and Patterson, Turing Lecture 2018, overlaid over "42 Years of Processors Data" https://www.karirupp.net/2018/02/42-years-of-microprocessor-trend-data/; "First Wave" added by Les Wilson, Frank Schirrmeister Original data up to the year 2010 collected and plotted by M. Horowitz, F. Labonte, O. Shacham, K. Olukotun, L. Hammond, and C. Batten New plot and data collected for 2010-2017 by K. Rupp

# SUPERCOMPUTING

## SUPERCOMPUTING EVOLUTION

Architecture paradigms

- 1940 1950: first computers are Supercomputers (specialized, expensive)
- 1960 1980: vector computers dominate HPC, while general purpose computers come to market at much lower prices
  - Focus: floating operations (linear Algebra)



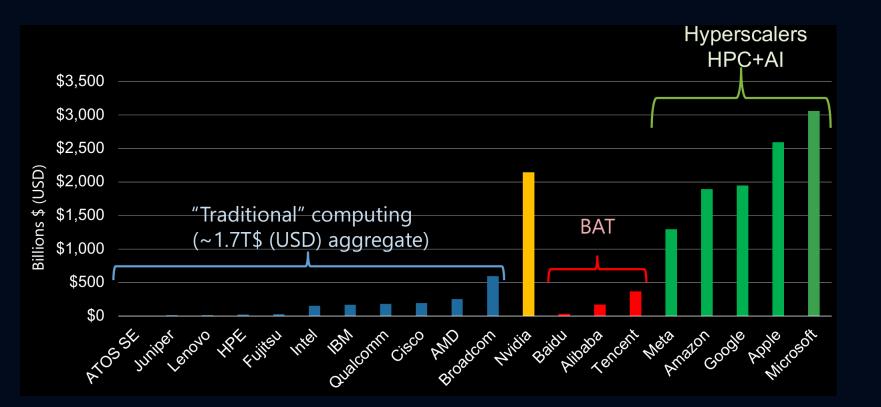
- Special purpose technologies (fast vector processors, parallel architectures)
- Only few machines produced  $\rightarrow$  expensive!
- 1990 2000: cluster computers are born
  - Integrate general purpose CPUs in HPC → more economic approach
  - Many "computers" connected through fast network
  - Distributed memory 
     → MPI
- 2010 2020: heterogeneous cluster systems
  - − CPU + Accelerator technologies (mostly GPUs)  $\rightarrow$  more FLOPS/Watt
  - Intel / AMD + NVIDIA / AMD / Intel
- 2020 today: very large GPU-based systems in HPC, while hyperscalers dominate AI-market, drive GPU design (and price), and build their own processors for their clouds







#### **MARKET CAPITALIZATIONS**



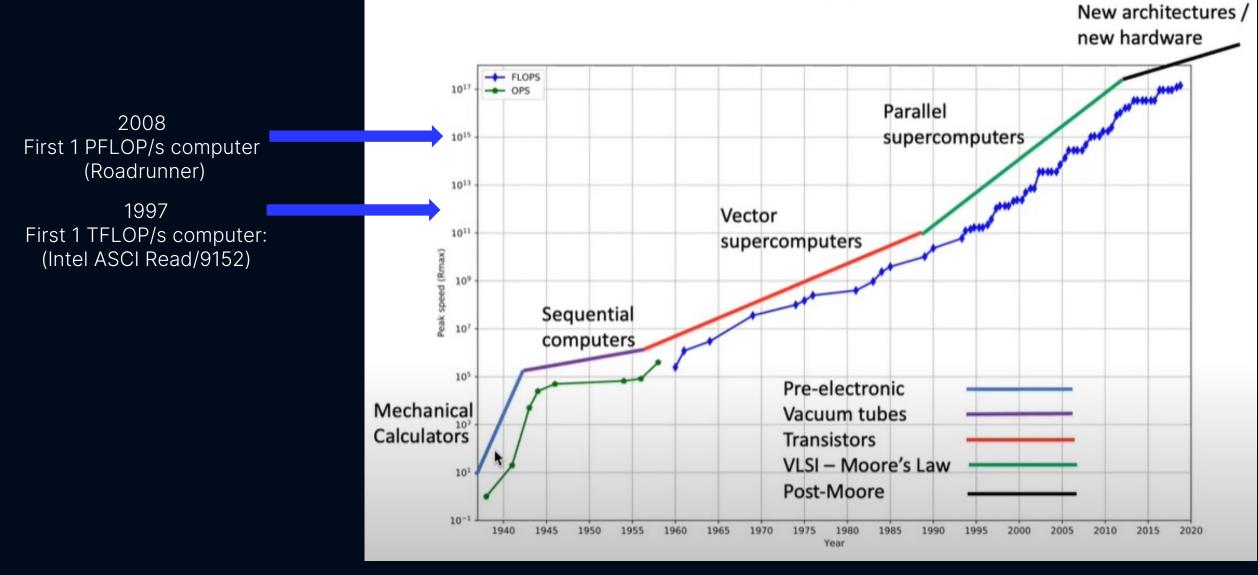
#### HPC looks at Cloud Computing

- Availability
- Cost vs price
- Higher level programming

Source: Reed, Gannon, Dongarra

Katherine Yelic, "Beyond Exascale Computing", ISC High Performance 2024

#### PERFORMANCE OF THE WORLD'S FASTEST SUPERCOMPUTERS



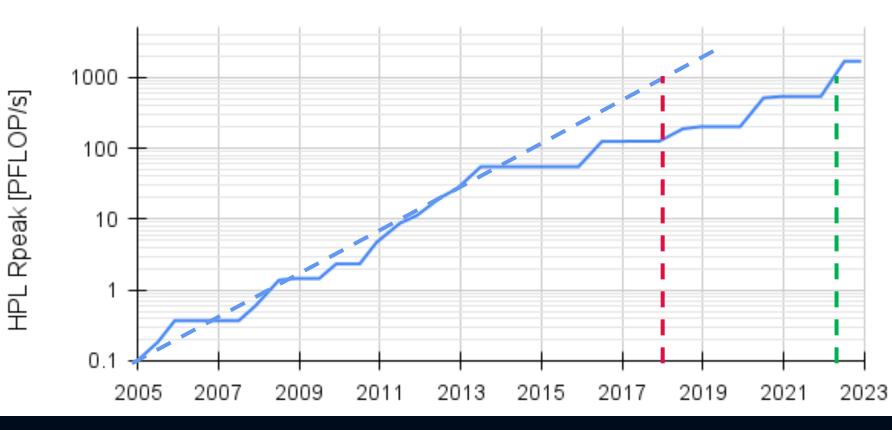
\*floating point operations per second (FLOPS, flops or flop/s)

Rob Schreiber, High Performance Computing: Beyond Moore's Law https://www.youtube.com/watch?v=LOf57fdxIn4&t=258s

# FROM PETASCALE TO EXASCALE COMPUTING

#### **EXASCALE ERA**

#### Top #1: HPL Rpeak [PFLOP/s]



- 1997: First 1 TFLOP/s computer: (Intel ASCI Read/9152)
- 2008: First 1 PFLOP/s computer: (*Roadrunner*)
- So.... First 1 EFLOP/s computer: 2018 !!
  - Well... not really
- It took 4 more years... 2022

https://www.top500.org/

#### FRONTIER First Exascale System

https://www.ornl.gov/news/ornl-celebrates-launch-frontier-worlds-fastest-supercomputer



Credit: Estela Suarez (Jülich Supercomputing Centre

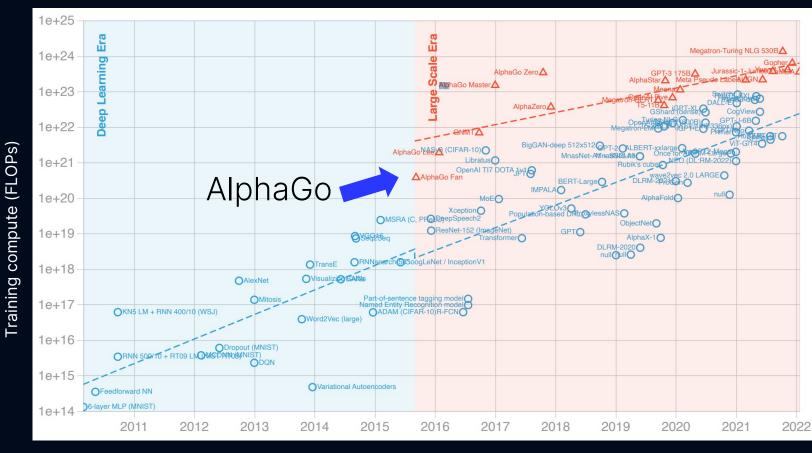
# OUTLINE

**1.** Evolution of Computing

#### 2. Machine Learning drives computational demands

- 3. Supercomputing and Parallel Programming
- 4. Jülich Supercomputing Centre

## THE ERA OF LARGE-SCALE DEEP LEARNING



• 2015: a new trend of large-scale models

- Computational capacity significantly higher (e.g., AlphaGo) than other models published in the same year
- OpenAl estimates 3.4-month doubling!

Publication date

## **NEW TRENDS IN AI FOUNDATION MODELS**

Strong trend towards FMs trained on extensive domain-agnostic datasets, using:

- Unsupervised learning
- Self-supervised representation learning,
- Multimodal learning

**Chat GTP-4 Could Pass the Bar** 

Exam

How Our Technology Evolves FAST

Brian Lamacraft · Follow

Published in ILLUMINATION . 2 min read . Jan 5

Deliver more robust insights and decision-making, and bring advances in:

- Mainstream problems, e.g.: Natural Language Processing (NLP), Computer Vision
- But also to many scientific fields, e.g., Earth observation

## AI chatbot's MBA exam pass poses test for business schools

 $\label{eq:chatGPT} ChatGPT \ earned \ a \ solid \ grade \ and \ outperformed \ some \ humans \ on \ a \ Wharton \ course$ 

#### ChatGPT Passes Google Coding Interview for Level 3 Engineer With \$183K Salary

'Amazingly, ChatGPT gets hired at L3 when interviewed for a coding position,' reads a Google document, but ChatGPT itself says it can't replicate human creativity and problem-solving skills.

By Emily Dreibelbis February 1, 2023 🕇 🎔 🖬 🚥

#### AI Passes U.S. Medical Licensing Exam

- Two papers show that large language models, including ChatGPT, can pass the USMLE

by Michael DePeau-Wilson, Enterprise & Investigative Writer, MedPage Today January 19, 2023

#### **CURRENT POPULARITY OF SUPERCOMPUTERS**

#### TPU v4: An Optically Reconfigurable Supercomputer for Machine Learning with Hardware Support for Embeddings Industrial Product\*

Norman P. Jouppi, George Kurian, Sheng Li, Peter Ma, Rahul Nagarajan, Lifeng Nai, Nishant Patil, Suvinay Subramanian, Andy Swing, Brian Towles, Cliff Young, Xiang Zhou, Zongwei Zhou, and David Patterson Google, Mountain View, CA



Tech > Science BABY STEPS Google artificial intelligence supercomputer creates its own 'Al child' that can outperform its human-made rivals The NASNet system was created by a neural network called AutoML earlier this year <u>Mark Hodge</u> Published: 1522, 5 Dec 2017 | Updated: 1127, 6 Dec 2017

#### RESEARCH Introducing the AI Research SuperCluster —

Meta's cutting-edge AI supercomputer for AI research

- Big tech companies deploy their AI supercomputers
- Supercomputing now goes far beyond traditional scientific computing, which was driven by large governments
- Major industries building highly specialized supercomputers are taking the lead

N. P. Jouppi, G. Kurian, et al., "TPU v4: An Optically Reconfigurable Supercomputer for Machine Learning with Hardware Support for Embeddings", 2023, https://doi.org/10.48550/arXiv.2304.01433

https://www.forbes.com/sites/jamesmorris/2022/10/06/teslas-biggest-news-at-ai-day-was-the-dojo-supercomputer-not-the-optimus-robot/?sh=22ba4ab780bd

https://ai.facebook.com/blog/ai-rsc/

https://www.thesun.co.uk/tech/5072741/google-nasnet-ai-child-reinforcement-learning/

Torsten Hoefler, "Efficient AI: From supercomputers to smartphones", Scalable Parallel Computing Lab @ ETH Zurich, https://youtu.be/xxwT45ljG4c

## HOW TO CREATE A FOUNDATION MODEL?

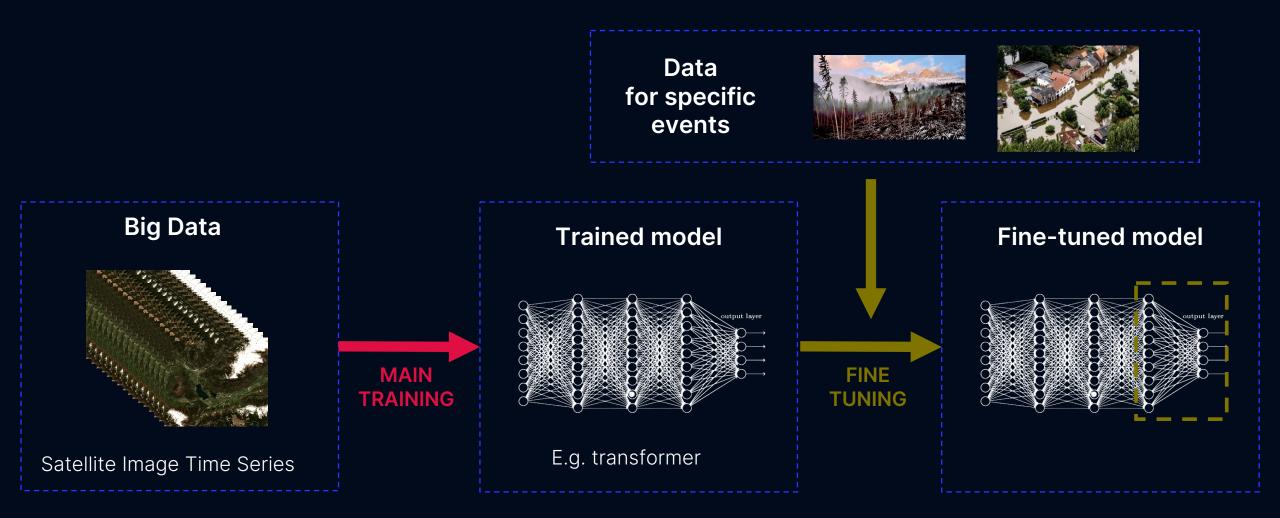
1) Gather data at scale

2) Train foundation model one time and evaluate

3) Fine-tune model for multiple downstream tasks

4) Inference (operational)

## HOW TO CREATE A FOUNDATION MODEL?



High Performance Computing

Cloud Computing, Computing workstations

#### TIME REQUIRED FOR A FULL TRAINING OF GPT-3 MODEL

175 billion weight parameters



Ranking in November 2020 (TOP500 (7 World, 1 Europe), Green500 (1 in TOP100) TOP10 AI (4)







- 1× Nvidia Ampere100 ≈ 90 years
- I× Nvidia Hopper100 ≈ 15-30 years
- 2,000× Nvidia Ampere100 ≈ 16 days (if scaled well on JUWELS Booster)

Julich Supercomputing Centre, "JUWELS Cluster and Booster: Exascale Pathfinder with Modular Supercomputing Architecture at Julich Supercomputing Centre," Journal of large-scale research facilities, vol. 7, no. A138, 2021.

S. Kesselheim, A. Herten, K. Krajsek, J. Ebert, J. Jitsev, M. Cherti, M. Langguth, B. Gong, S. Stadtler, A. Mozaffari, G. Cavallaro, R. Sedona, A. Schug, A. Strube, R. Kamath, M. G. Schultz, M. Riedel, and T. Lippert, "JUWELS Booster – A Supercomputer for Large-Scale AI Research," in High Performance Computing (H. Jagode, H. Anzt, H. Ltaief, and P. Luszczek, eds.), (Cham), pp. 453–468, Springer International Publishing, 2021.

Jenia Jitsev, Towards Scalable Deep Learning, Scalable Learning & Multi-Purpose AI Lab, Helmholtz AI, LAION @ JSC

#### TIME REQUIRED FOR A FULL TRAINING OF GPT-4 MODEL

1,8 Trillion weight parameters



Ranking in November 2020 (TOP500 (7 World, 1 Europe), Green500 (1 in TOP100) TOP10 AI (4)







- I× Nvidia Ampere100 ≈ 1,200 years
- I× Nvidia Hopper100 ≈ 200-600 years
- 2,000× Nvidia Ampere100 ≈ 900 days

Julich Supercomputing Centre, "JUWELS Cluster and Booster: Exascale Pathfinder with Modular Supercomputing Architecture at Julich Supercomputing Centre," Journal of large-scale research facilities, vol. 7, no. A138, 2021.

S. Kesselheim, A. Herten, K. Krajsek, J. Ebert, J. Jitsev, M. Cherti, M. Langguth, B. Gong, S. Stadtler, A. Mozaffari, G. Cavallaro, R. Sedona, A. Schug, A. Strube, R. Kamath, M. G. Schultz, M. Riedel, and T. Lippert, "JUWELS Booster – A Supercomputer for Large-Scale Al Research," in High Performance Computing (H. Jagode, H. Anzt, H. Ltaief, and P. Luszczek, eds.), (Cham), pp. 453–468, Springer International Publishing, 2021.

Jenia Jitsev, Towards Scalable Deep Learning, Scalable Learning & Multi-Purpose AI Lab, Helmholtz AI, LAION @ JSC

# OUTLINE

- **1. Evolution of Computing**
- 2. Machine Learning drives computational demands

#### 3. Supercomputing and Parallel Programming

4. Jülich Supercomputing Centre

# WHAT ACTUALLY IS A SUPERCOMPUTER?

## WHY USE PARALLEL COMPUTERS?

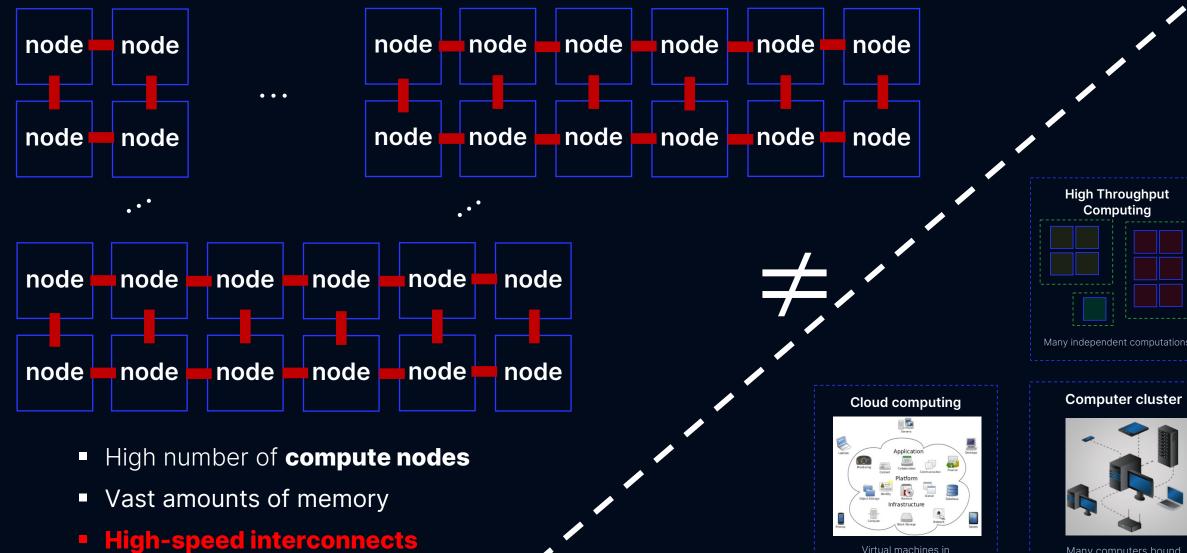
- Parallel computers can be the only way to achieve specific computational goals in a given time
- Sequential system is too "slow"
  - Calculation takes days, weeks, months, years, ...
  - Use more than one processor to get calculation faster
- Sequential system is too "small"
  - Data does not fit into the memory
  - Use parallel system to get access to more memory



https://9gag.com/gag/av5vmzd

• You realize you have a parallel system (**multicore**) and you want to make use of its special features

## **HIGH-PERFORMANCE COMPUTING SYSTEMS**



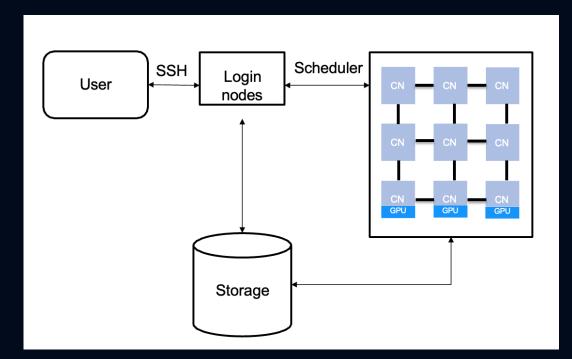
Many computers bound together locally

Virtual machines in

compute center(s

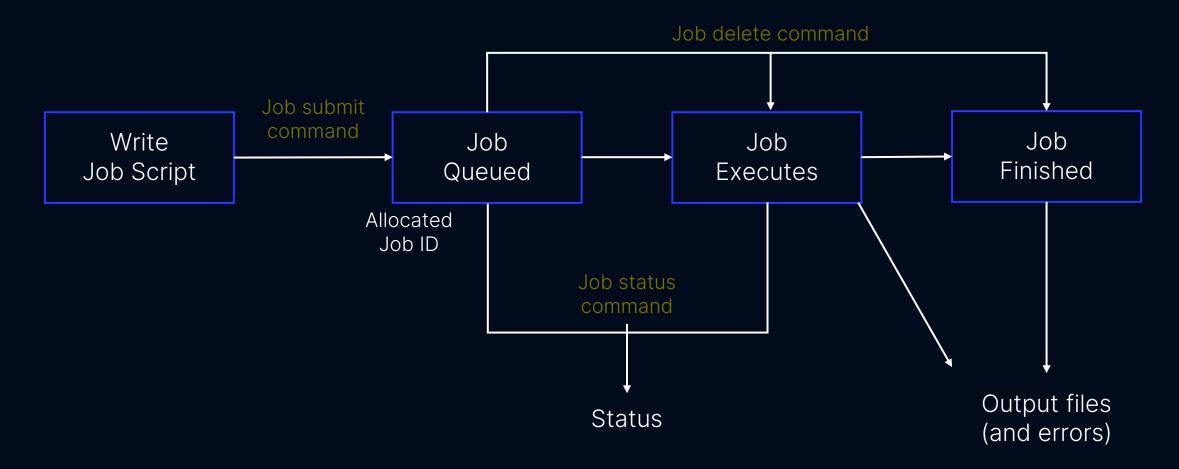
## **HPC BUILDING BLOCKS**

- Hardware
  - Login and compute nodes (CN)
  - Network
  - Storage
- Software
  - Operating System (OS)
  - Compilers
  - Libraries
  - Scheduler

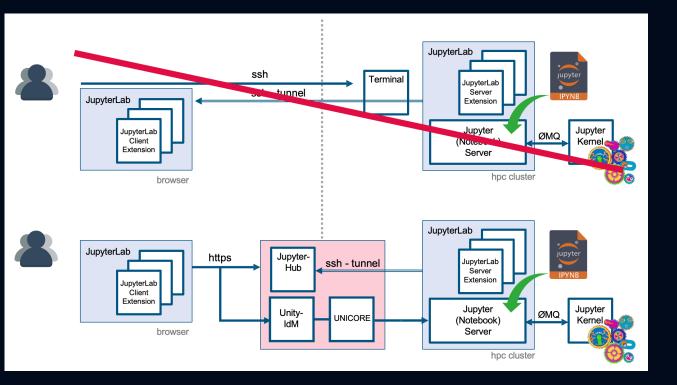


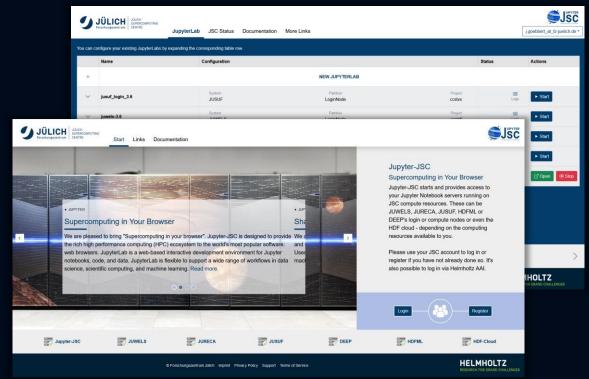
## USING THE SUPERCOMPUTER MEANS SUBMITTING A JOB TO A BATCH SYSTEM

Job scheduling according to priorities. The jobs with the highest priorities will be scheduled next.



#### JUPYTER-JSC WEBSERVICE

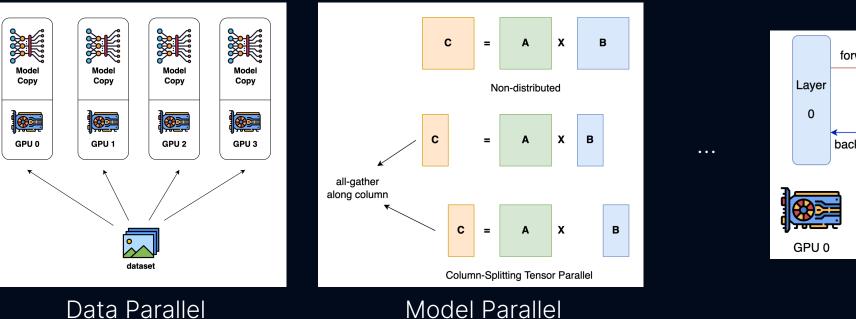


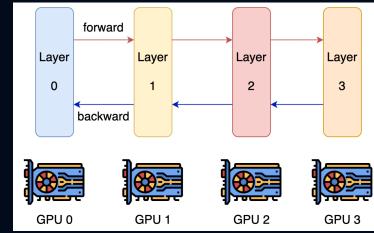


Credit: Jens H. Göbbert Tim Kreuzer, (Jülich Supercomputing Centre) https://jupyter.jsc.fz-juelich.de/

## DISTRIBUTED DEEP LEARNING

Paradigms of Parallelism





#### **Pipeline Parallel**

Credit: Colossal-Al, https://colossalai.org/docs/concepts/paradigms\_of\_parallelism/

# OUTLINE

- **1. Evolution of Computing**
- 2. Machine Learning drives computational demands
- **3.** Supercomputing and Parallel Programming

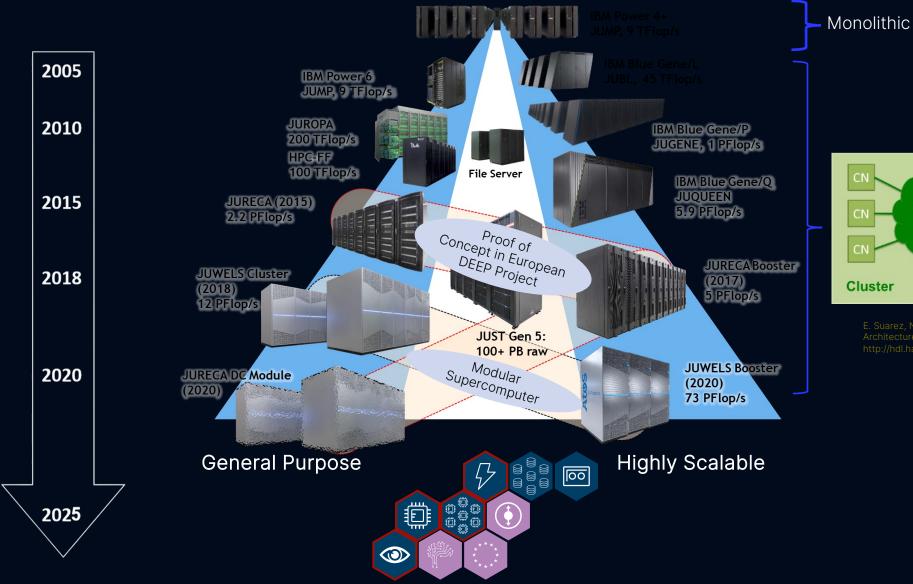
#### **4. Jülich Supercomputing Centre**

## JÜLICH SUPERCOMPUTING CENTRE (JSC)

Multi-system facility - Tier-0/1 HPC resources of the highest performance class



## (DUAL) HARDWARE STRATEGY AT JSC

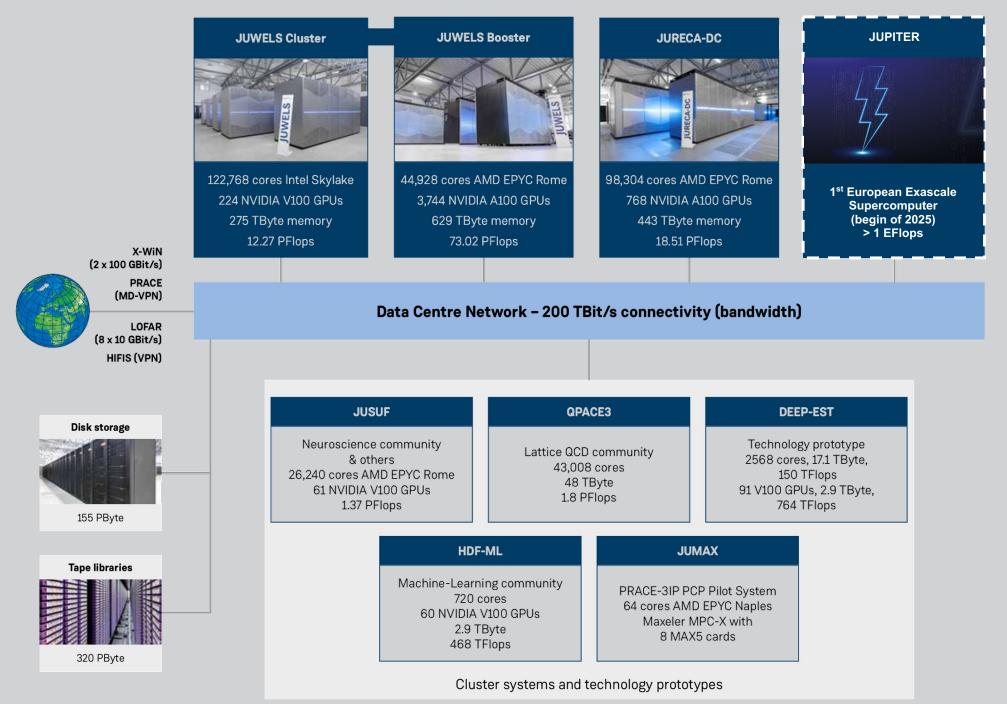


> E. Suarez, N. Eicker, and T. Lippert, "Modular Supercomputing Architecture: from Idea to Production", 2019 http://hdl.handle.net/2128/22212



https://www.deep-projects.eu

Exascale Modular Supercomputer



Credit: Jens Henrik Göbbert (Jülich Supercomputing Centre)



- Intended for mixed capacity and capability workloads
- Designed with big-data science needs in mind





JURECA-DC, Jülich Supercomputing Centre (JSC), https://www.fz-juelich.de/en/ias/jsc/systems/supercomputers/jureca

#### **JURECA-DC CPU NODES**

- 576 compute nodes Atos Atos
  - 2× 64-core AMD Epyc 7742 Rome CPUs AMD& AMD
    - 2x 8 memory channels
    - 2x 256 GB DDR4 @ 3.2 GHz
      - > 96 nodes with 2x 512 GB DDR4 @ 3.2 GHz
    - $_{\circ}$  2x 4 NUMA domains
  - PCle Gen4 1x HDR100 InfiniBand adapter (100Gbps) 💿 IVIDIA



#### **JURECA-DC GPU NODES**

- 192 compute nodes Atos
  - 2× 64-core AMD Epyc 7742 Rome CPUs AMD
    - 2x 8 memory channels
    - 。 2x 256 GB DDR4 @ 3.2GHz
    - $_{\circ}$  96 PCIe Gen4 lanes
  - 512 GB DDR memory
  - 4x Nvidia A100 GPUs 👁 🛯 🕬 🗠 🗠 🗠 🗠 4x Nvidia A100 GPUs
    - $_{\circ}$   $\,$  9.7 / 19.5 TF/s peak
    - $_{\circ}$  40 GB HBM2
    - 1.5 TB/s memory performance
    - 。 NVLink3 full mesh
      - 4 links (200GB/s) between GPU pairs
  - PCle Gen4 x32 (64 GB/s)
- 2x HDR200 InfiniBand adapter (1 per GPU) InfiniBand adapter (1 per GPU)





## JURECA-DC LOGIN NODES

- 12 login nodes
  - 2x 64-core AMD Epyc 7742 Rome CPUs
  - 1024 GB DDR4 @ 3.2 GHz
  - 100 GigE external network
  - 2x Nvidia RTX8000 GPUs
    - Different compute capabilities than in compute nodes!
  - Used for:
    - Compile/submit jobs
      - Careful with make -j !
    - Small pre- and post-processing/visualization
  - Shared nodes!



## JUPITER

#### The Arrival of Exascale in Europe

- JUPITER: EuroHPC JU system hosted at JSC
- Launched with focus on applications
- ~6000 nodes,
   24 000 H100 GPUs, 1 728 000 Arm cores, 24 000 NDR200 endpoints
- Landing in Modular Data Center
- Preparation is ongoing:
  - JUREAP
  - GH200 test systems
- $\rightarrow$  jupiter.fz-juelich.de







with 72.733 GFlops/watts Performance in the Green500 List published at the ISC24 Conference on June 01, 2024. Congratulations from the Green500 Editors







#### Funding Agencies





uroHPC



Ministerium für Innovation, Wissenschaft und Forschung des Landes Nordrhein-Westfalen



## Thank you for your attention