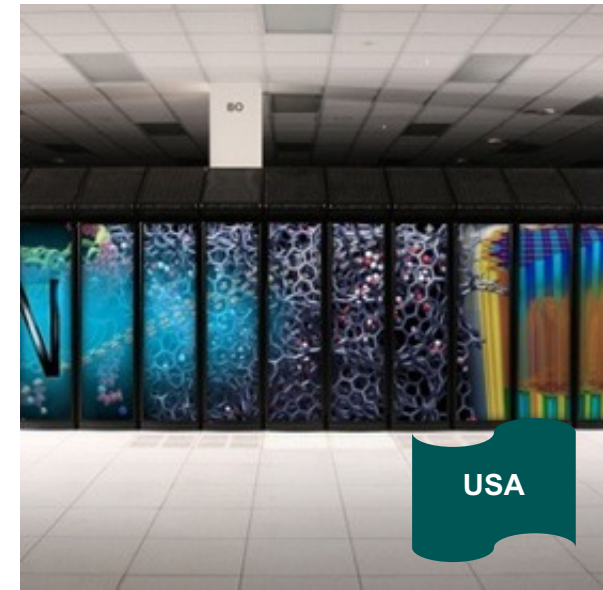
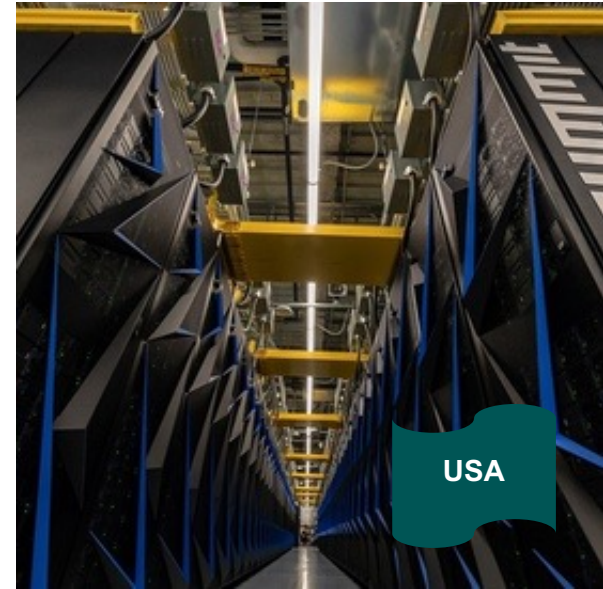
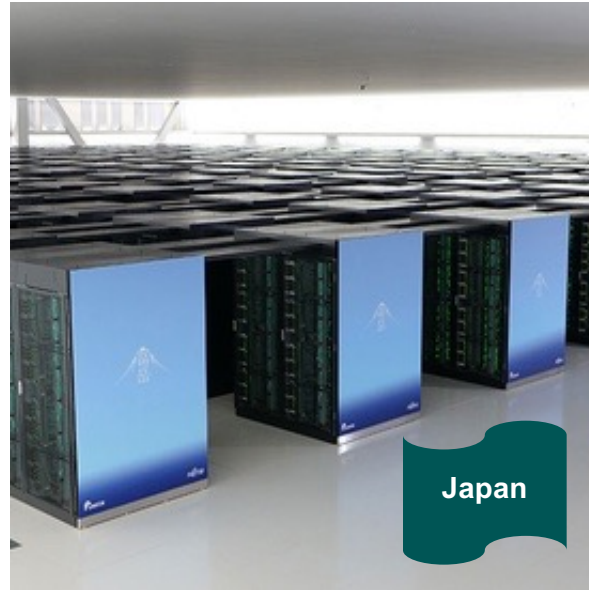




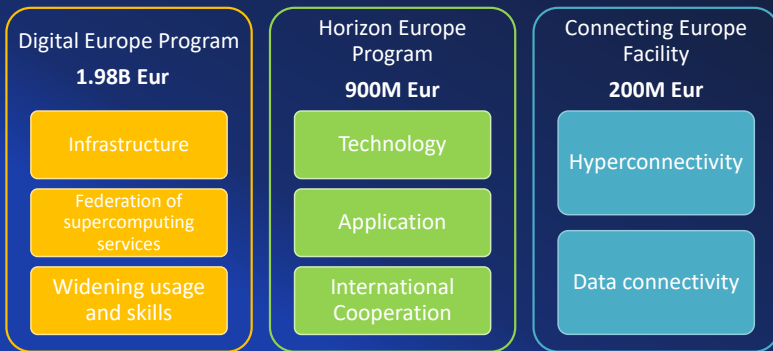
**HPC (AND CLOUD) IN EUROPE AND AT THE  
MAX PLANCK SOCIETY**

**Erwin Laure, Director MPCDF**  
**Markus Rampp, Deputy Director MPCDF**  
**Salishan Conference, 2023-04-25**



# The EuroHPC Joint Undertaking (since 2018)

## LEVEL AND SOURCES OF EU FUNDING 2021-2027



\*Member states to match this with national contributions

**Mission:** Establish an integrated world-class supercomputing & data infrastructure and support a highly competitive and innovative HPC and Big Data ecosystem

## OUR MEMBERS

- 31 participating countries
- The European Union (represented by the European Commission)
- Private partners



Infrastructure & Operations

R&I, Applications & Skills

HPC Ecosystem



# EUROHPC SYSTEMS

- 1 more pre-exascale @ BSC, 2023
- “Jupiter”@Jülich will be the first European Exascale system (500 M€), 2024





## EUROPEAN PRE-EXASCALE SYSTEMS



### LUMI @ CSC (Finland): HPE/CRAY

- AMD EPYC CPUs, AMD MI250 GPUs
- 2.2 M CPU cores & GPU compute units
- 309 PFlop/s HPL, rank 3 in Top500 11/2022

### LEONARDO @ CINECA (Italy): Atos

- Intel Xeon CPUs, Nvidia A100 GPUs
- 1.5 M CPU cores & GPU streaming multiprocessors)
- 175 PFlop/s HPL, rank 4 in Top500 11/2022





## APPLICATION DEVELOPMENT

### Centres of Excellence (CoEs)

- **Improve applications important to certain domains towards the Exascale**
  - Optimization, new algorithms, methods, etc.
- **Provide training and support**
- **Cover full workflow (including data handling)**
  
- **Covering a wide range of scientific domains**

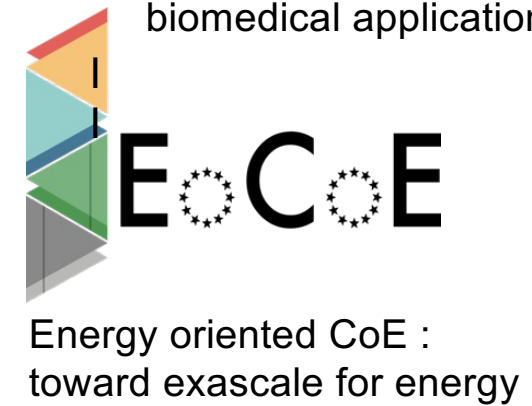
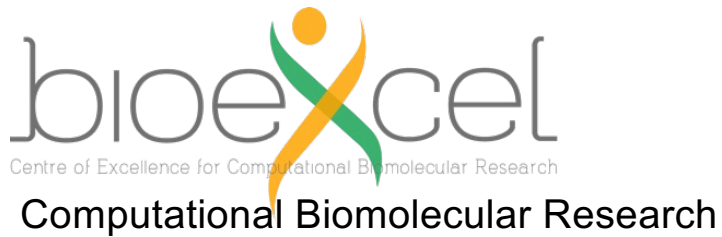


<https://www.hpccoe.eu/>



# FIRST 15 CENTRES OF EXCELLENCE

15 Centres of excellence active up to 2022: created during three calls (2015, 2018 and 2019)





## FIRST 15 CENTRES OF EXCELLENCE



**esiwace**  
CENTRE OF EXCELLENCE IN SIMULATION OF WEATHER  
AND CLIMATE IN EUROPE



**HiDALGO**

HPC and Big Data Technologies  
for Global Challenges



**EXCELLERAT**

CoE for Engineering Applications



Materials design at the exascale



NOVEL MATERIALS DISCOVERY



Performance Optimisation and Productivity



Targeting Real chemical accuracy at the EXascale





## COES STARTED ON JAN 1, 2023





# COE AREAS OF EXPERTISE

CoEs develop strong expertise in their specific application fields as well as more transversal HPC skills needed to achieve science at exascale.

## Disciplinary Expertise

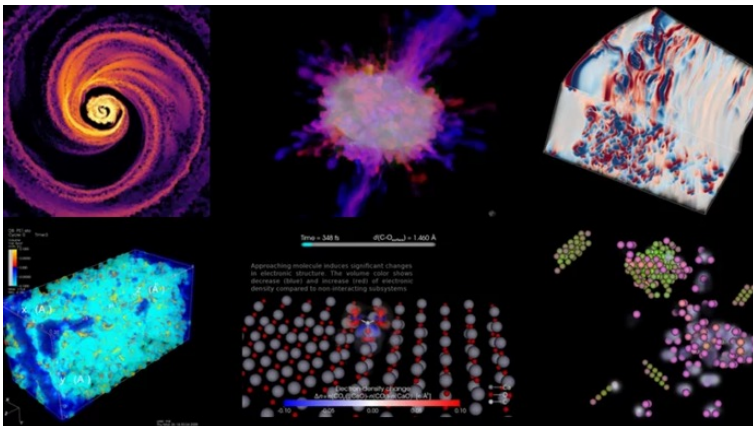
- Energy production (wind, hydro, fusion,...)
- Engineering (automotive, aerospace,...)
- Combustion
- Plasma physics
- Material science
- Material for energy (batteries, PV cells,...)
- Chemistry
- Climate sciences and weather forecasts
- Global challenges (health-relevant social habits, green growth, dynamics of global urbanisation.)
- Solid earth physics
- Molecular biology
- Personalized medicine
- Biomedical applications (Cardiovascular Medicine, Neuro Musculoskeletal Medicine,...)
- Astro physics

## HPC expertise

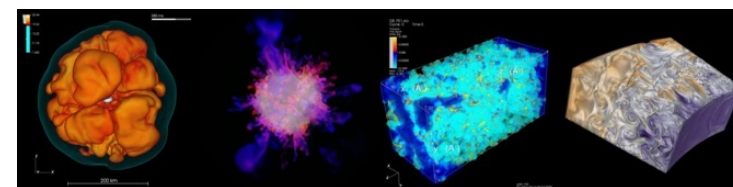
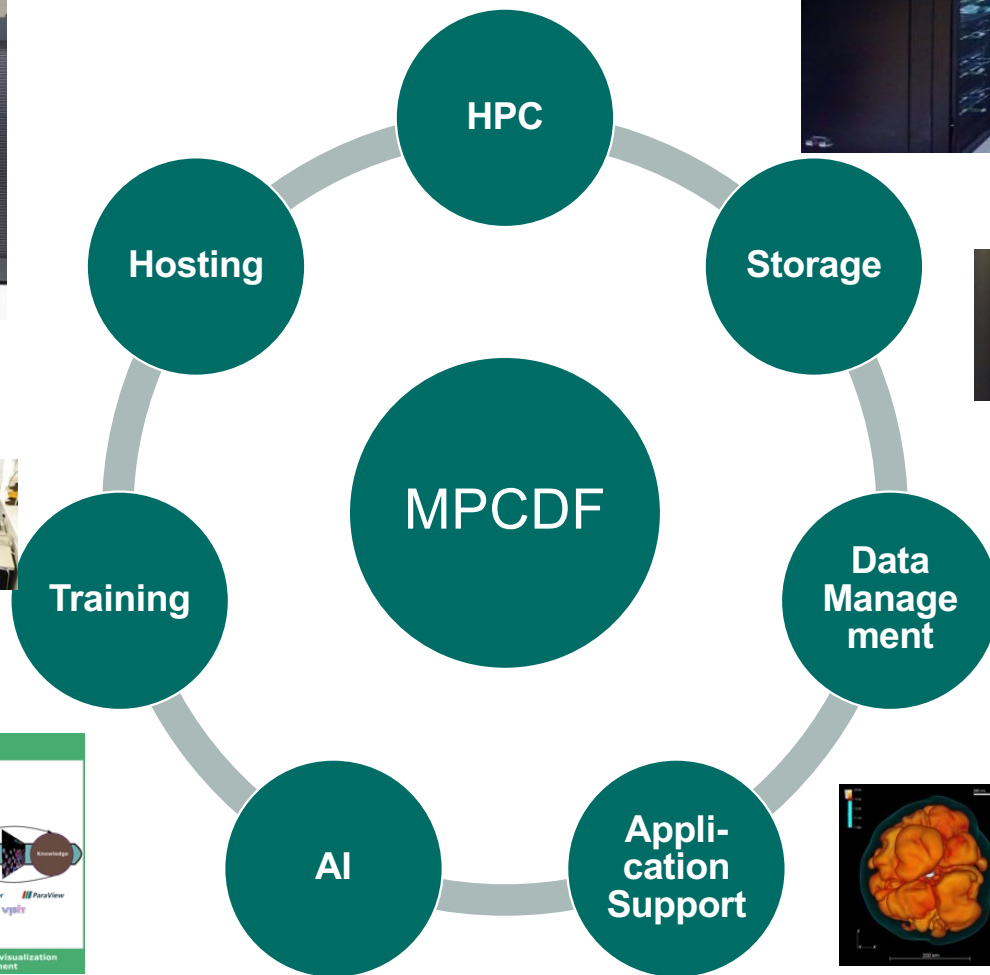
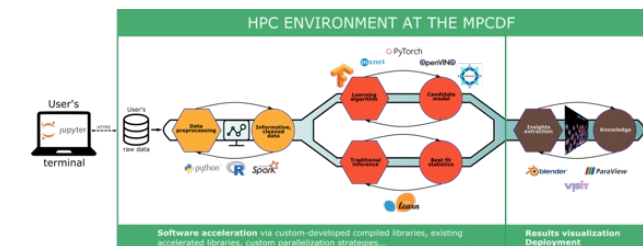
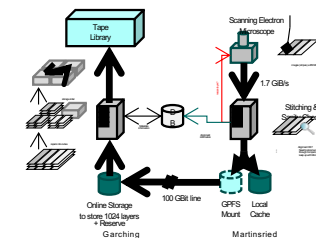
- Programming models for exascale
- Performance monitoring, optimization and scalability
- Tools for HPDA in complex workflows
- Workflows
- Scalable solvers, linear algebra
- Data flow, in-situ data analysis and I/O
- Ensemble runs
- Implementing co-design and technology integration



# MAX PLANCK COMPUTING AND DATA FACILITY (MPCDF): A CROSS-INSTITUTIONAL COMPETENCE CENTRE OF THE MAX PLANCK SOCIETY TO SUPPORT COMPUTATIONAL AND DATA SCIENCES



# CORE SERVICES OF THE MPCDF



# HPC RESOURCES FOR THE MPG

## Cobra (2018 – )

~ 12 PetaFlop/s aggregated peak performance

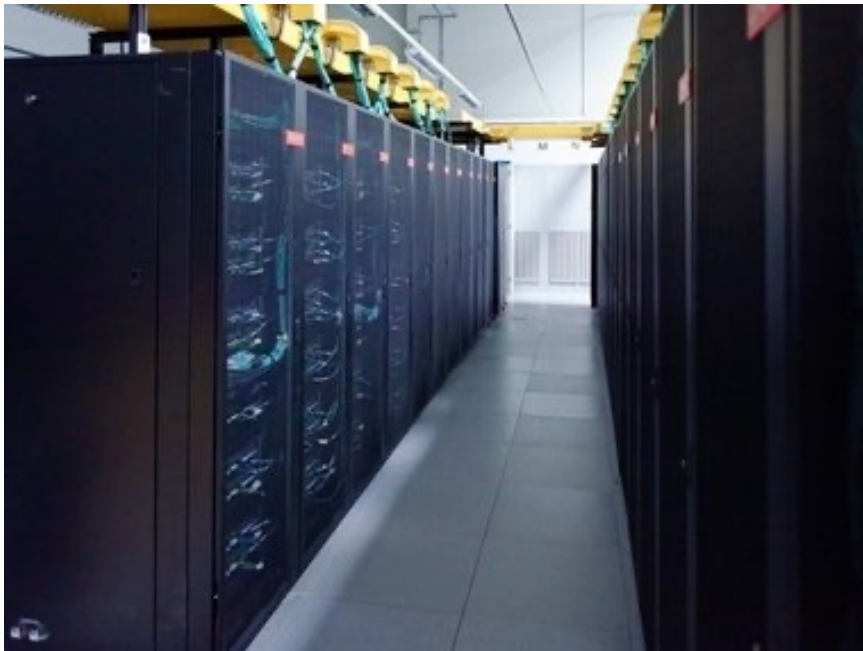
~ 3.400 compute nodes,

~ 137.000 compute cores (Intel Xeon Gold 6148 “Skylake-SP”)

128 NVIDIA Tesla V100 GPUs; 240 Quadro RTX 5000 GPUs

OmniPath 100 Gb/s Interconnect

Rank 96 in June’22 TOP500 list



MAX PLANCK COMPUTING AND DATA FACILITY, ERWIN LAURE & MARKUS RAMPP



## Raven & Raven GPU (2020/2021 – )

~ 5 PetaFlop/s (CPU), ~15 PetaFlop/s (GPU)

~ 1.600 compute nodes (CPU),

~ 115.000 compute cores (Intel Xeon Platinum 8360Y “Icelake-SP”)

192 GPU nodes with **768 NVIDIA A100 GPUs**

HDR Infiniband Interconnect (100 Gb/s ... 400 Gb/s)

Rank 61 and 99 in June’22 TOP500 list

# NEW SYSTEM 2023/24 ...



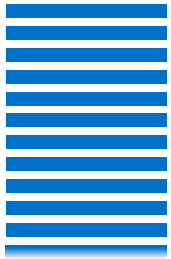
## Genoa

## MI300



# HPC-CLOUD

## Cloud Compute



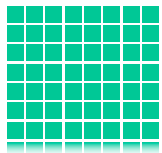
- Xeon CPUs
- Nvidia A30/A100
- 512-4096 GB RAM
- SSD/NVMe



700 Cores  
70 GPUs  
140 TB RAM



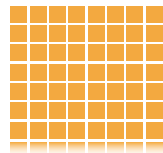
## Block & Object Storage



SSD Pool: 80 TB  
HDD Pool: 6 PB



## File System



HDD Pool: 3.5 PB



**Nexus**





# HPC-CLOUD CONCEPT

*A general solution for complex workflows, complementing the HPC systems*

## *HPC Cloud*

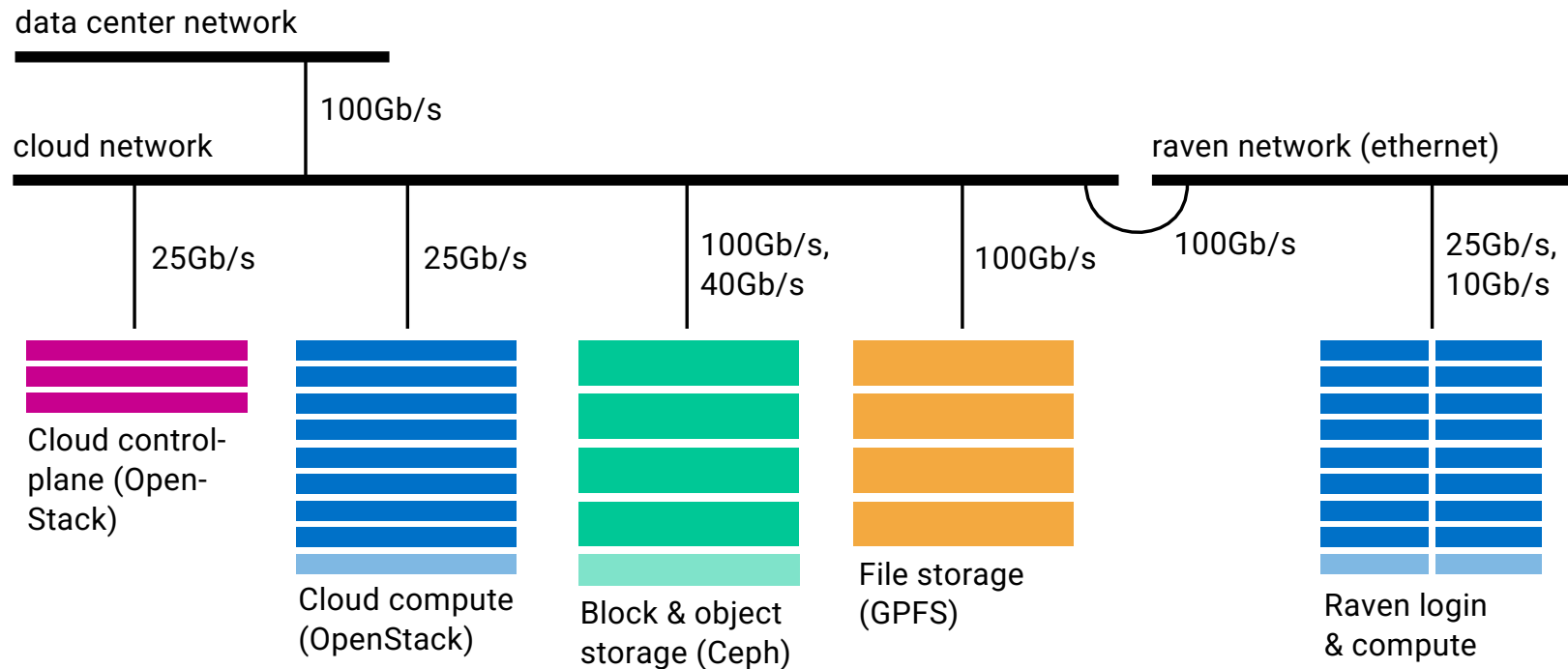
- ✓ Logically positioned near super-computing resources, esp. Raven
- ✓ Contains significant compute power
- ✗ Not itself a parallel compute cluster

- ✓ Flexible computing environment
- ✓ Infrastructure-as-a-Service inc. self-service dashboard, standard APIs
- ✗ Not intended for general IT services





# NETWORK ARCHITECTURE

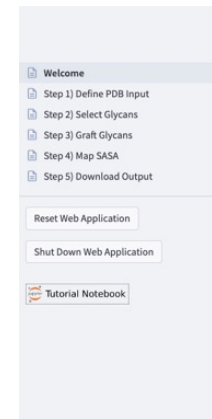


<https://docs.mpcdf.mpg.de/doc/cloud/index.html>



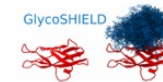
# HPC CLOUD USECASES

- Online Data Analysis
- Training (BinderHub)
- Data Publication
- Image Processing
- Long-running jobs



MAX PLANCK INSTITUTE OF BIOPHYSICS | Inserm | MAX PLANCK COMPUTING & DATA FACILITY

## Welcome to the GlycoSHIELD web application!

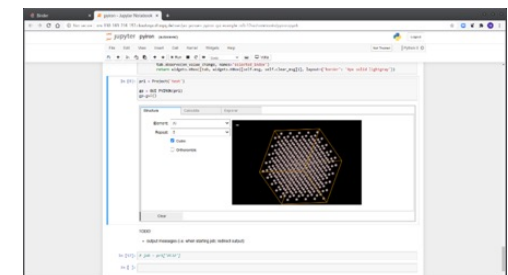
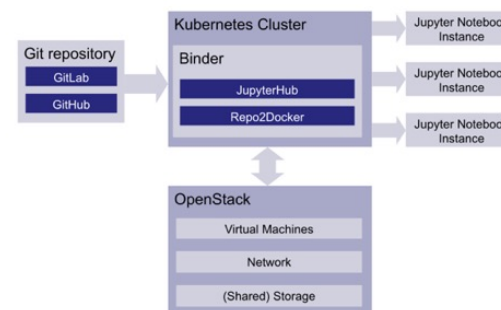
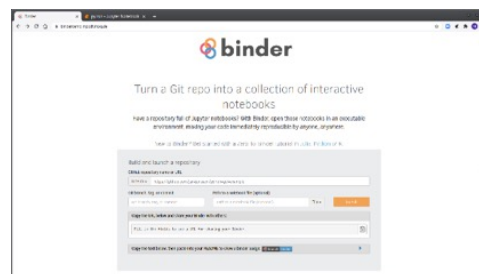


This application enables users to graft glycan conformer arrays on specific amino-acids of a selected protein structure. 50 glycan types are currently available in the library.

To use the application, please navigate from step 1 to step 5 using the sidebar on the left and follow the instructions provided at each step. (Mobile devices: use arrow to show the side bar)

This application was developed as an open-access community resource and is maintained by Klaus Reuter, Matt Sikora and Cyril Hanus. Please support us by citing our work: <https://doi.org/10.1101/2021.08.04.455134>

The GlycoSHIELD source code as well as convenience scripts are available on [gitlab](#)



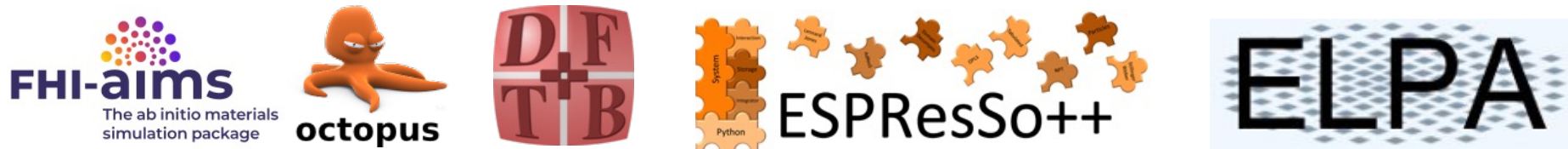


# HPC APPLICATION SUPPORT FOR THE MPG

Original contributions and long-term support for

*development, optimization and porting of HPC and AI applications*

FHI-aims, OCTOPUS, NECI, DFTB+, ESPResSo++ (*materials science*), ELPA (*eigensolver library*)



GENE, SeLaLib, IDE, TORBEAM, GRILLIX, TurTLE, MagIC, CHIEF (*plasma & astrophysics*)



BioEM, COMPLEXES++, GlycoSHIELD, TriMEM, GROMACS, DIAMOND, ... (*biophysics, bioinformatics*)

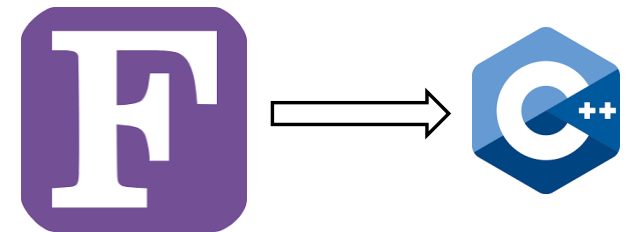




## APPLICATION SUPPORT: FOCUS TOPICS

### GPU porting and optimization of major production codes developed in the Max Planck Society:

- GENE, GRILLIX/PARALLAX (plasma physics / nuclear fusion research)
- OCTOPUS, FHI-aims (electronic structure theory / DFT, TD-DFT)
- ELPA (eigensolver library)
- DIAMOND (bioinformatics / fast sequence alignment)



### Portable programming models and frameworks (and related SW-engineering challenges):

- OpenMP, OpenACC, HIP, SYCL, ..., Kokkos

### Beyond „traditional“ MPI communication and I/O

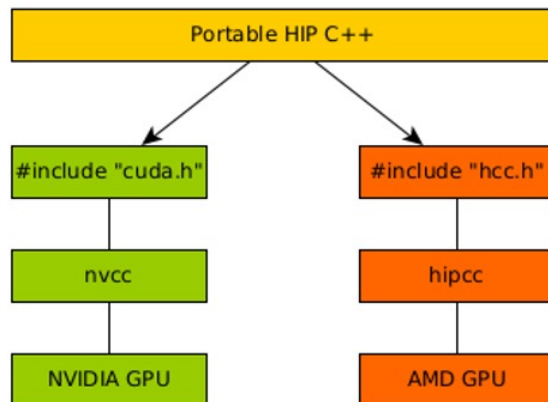
- asynchronous and GPU-aware MPI, HPX, in-situ techniques, burst-buffer filesystems



# AMD HIP: FROM NVIDIA (CUDA) TO AMD (HIP) AND BACK(?)

On the AMD platform, HIP is the native programming paradigm *and portability layer* for ROCm

- HIP enables compatibility between Nvidia and AMD GPUs
- HIP essentially adopts CUDA semantics



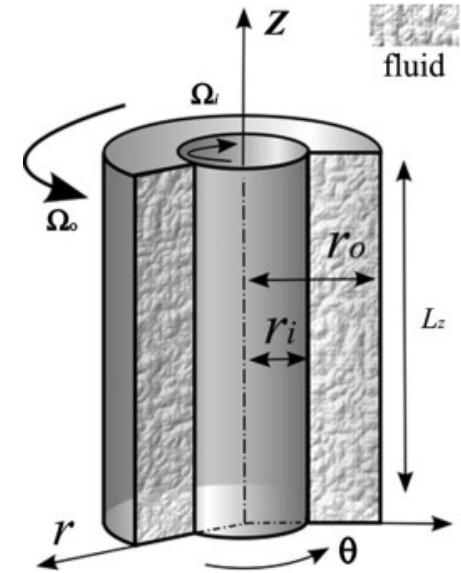
CUDA (Nvidia)	HIP (portable)	ROCm (AMD)
cuBLAS	hipBLAS	rocBLAS
cuFFT	hipFFT	rocFFT
cuRAND	hipRAND	rocRAND
cuSPARSE	hipSPARSE	rocSPARSE
NCCL		RCCL
CUB	hipCUB	rocPRIM

<https://www.lumi-supercomputer.eu/preparing-codes-for-lumi-converting-cuda-applications-to-hip/>



## MY FAVOURITE TEST VEHICLE: NSCOUETTE

- Pseudospectral DNS code in Taylor-Couette geometry:
  - Taylor-Couette devices, pipes, (astrophysical) discs, ...
  - FFTs, global transposes, linear systems
- Open Source:
  - <https://github.com/dfeldmann/nsCouette>
  - <https://gitlab.mpcdf.mpg.de/mjr/nscouette>
  - Jose M. Lopez, Daniel Feldmann, Markus Rampp, Alberto Vela-Martin, Liang Shi & Marc Avila, nsCouette - A high-performance code for direct numerical simulations of turbulent Taylor-Couette flow, *SoftwareX*, 11, 100395 (2020), preprint: arXiv:1908.00587v3
- Scalable MPI-OpenMP Fortran version for CPUs: *main workhorse*
- **Single-GPU CUDA version: development by A. Vela-Martin**





## BENCHMARKS WITH NSCOUETTE-GPU (GRID: 128,512,257)

Platform	GPU	FP64 peak [Tflop/s] <i>(ratio)</i>	BW peak [GB/s] <i>(ratio)</i>	Runtime CUDA code [ms/step] <i>(ratio)</i>	Runtime HIP code [ms/step] <i>(ratio)</i>
Nvidia	A100	9.7 <i>(1.00)</i>	1555 <i>(1.00)</i>	228.3 <i>(1.00)</i>	228.2 <i>(1.00)</i>
Nvidia	V100	7.0	900 <i>(0.58)</i>	388.5 <i>(0.59)</i>	(not tested)
AMD	MI210	22.6	1638 <i>(1.05)</i>	N/A	296.5 <i>(0.77)</i>

### Main observations:

- **smooth and automatic conversion CUDA->HIP with hippify tool**
- **software: zero HIP overhead on A100 (no surprise: hipcc -> nvcc, hipBLAS -> cuBLAS, ...)**
- **AMD hardware/software: relative underperformance on MI210 (0.77)**
- **AMD profiling tools are maturing**



# OPENMP FOR GPUS: BS-SOLCTRA

## BS-SOLCTRA: Biot-Savart Solver for Computing and Tracing Magnetic Field Lines

- C++ code developed by CeNAT (Costa-Rica) for the Stellarator of Costa Rica 1 (SCR1)
- trivially parallel CPU code (OpenMP + MPI) over particles
- deep call stack down to hotspot computations

Test Case	Average Total Execution Time [s]			
	Intel Xeon IceLake SP Node (72 cores, 1 thread per core)	NVIDIA A100 Prescriptive	NVIDIA A100 Descriptive	AMD MI250 (1 MCM)
Small	115.34	25.83	24.83	21.99
Medium	576.72	111.44	109.29	96.79
Large	1155.60	216.52	213.30	191.56

+ demonstrated portability to Intel Xe GPU (PonteVeccio preview)

- single-source not achieved here (data structures)

```
#pragma omp target enter data map (to:...)
for(int i=0;i<steps;i++){
#pragma omp target teams distribute parallel for
  for(int p=0; p<particle_count;p++)
    computeIteration(...)
}
}
#pragma omp target exit data map (from:...)
```

Implementing a GPU-Portable Field Line Tracing Application with OpenMP Offload, D. Jimenez, J. Herrera-Mora, M. Rampp, E. Laure, E. Menses, CARLA (2022)





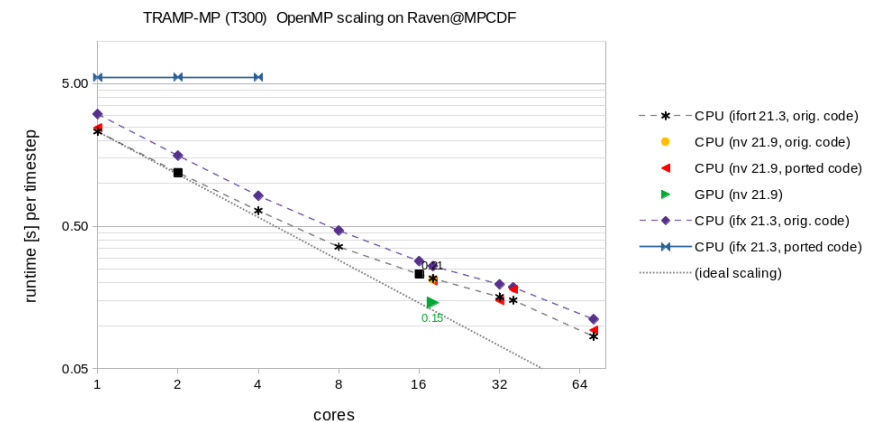
# OPENMP FOR GPUS: TRAMP

## TRAMP: 3D-radiation-hydrodynamics (FLD) code by MPI for Astronomy

- classic (F77 / OpenMP) legacy CPU code with 3-nested loop structure

```
!$OMP TARGET TEAMS LOOP COLLAPSE(2) REDUCTION(+:RHSNORM)
  DO K=KMIN,NZ
    DO J=1,NY
!$OMP LOOP REDUCTION(+:RHSNORM)
      DO I=IMIN,NX
```

- single-source (CPU-GPU) with no performance overhead (baseline: original OpenMP CPU code with ifort)
- speedup on GPU within expectations of ported code parts
- further porting mostly hampered by **tedious data locality handling** (would greatly benefit from fast unified memory -> MI300, GRACE-HOPPER)

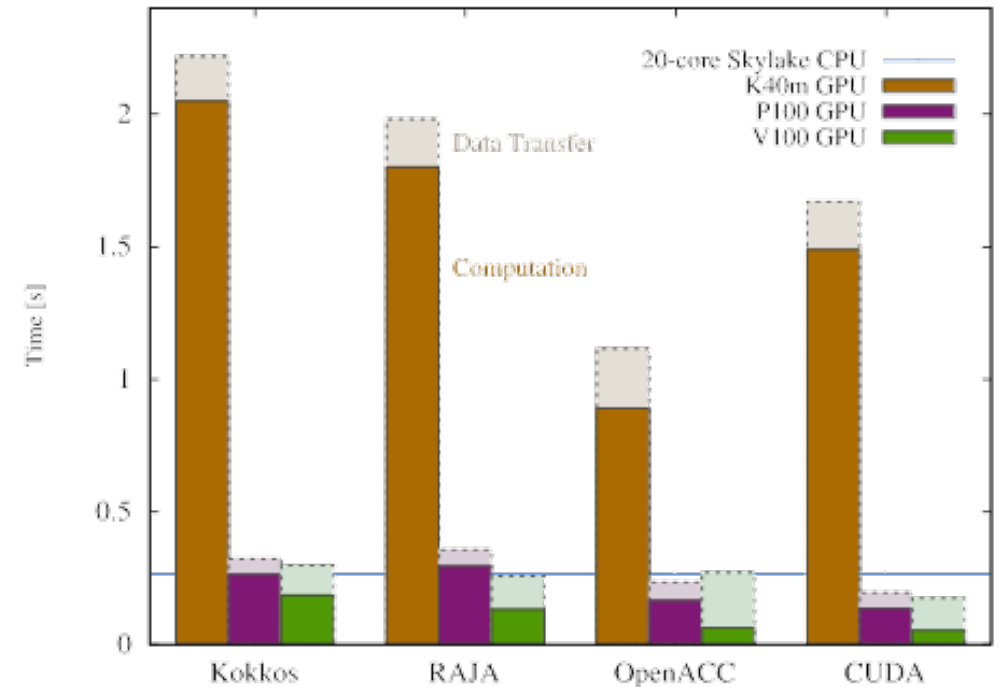




# KOKKOS: PORTABILITY STUDY & NEW VLASOV 6D CODE

- usability and performance study of high-level frameworks Kokkos and RAJA on GPUs and CPUs (NMPP & MPCDF, 2018)
- proxy application: non-trivial PIC kernel written in C++ (SeLaLib)
- Kokkos (and RAJA) appear mature and usable for our complex proxy application

=> Led to the development (from scratch) of a semi-lagrangian Vlasov Code in 6D with Kokkos:



N. Schild, M. Raeth, S. Eibl, K. Hallatschek, K. Kormann. A performance portable implementation of the semi-Lagrangian algorithm in six dimensions. arXiv: 2303.05994

V. Artigues, K. Kormann, M. Rampp, K. Reuter. Evaluation of performance portability frameworks for the implementation of a particle-in-cell code. *Concurrency and Computation: Practice and Experience* 32:e5640 (2020). arXiv:1911.08394

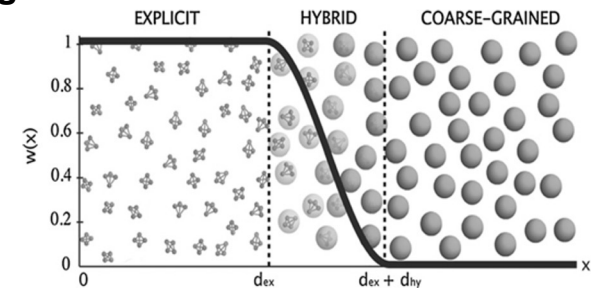


# KOKKOS: MOLECULAR DYNAMICS

In collaboration with Ch. Junghans (LANL)

## ESPRESSO++ (a classical molecular-dynamics package by MPI for Polymer Research)

- developed an adaptive-resolution MD code (H-AdResS) from scratch using Kokkos
- leveraging data-structures from Cabana (ECP), collaboration with LANL
- code is production-ready, promising GPU benchmarks:



Binary Lennard-Jones Mixture with 36000 particles

	CPU (MPUpS) 18 IceLake Cores	GPU (MPUpS) 1 A100	speedup (CPU -> GPU)
NPT	6.5	15.9	2.4x
NVT	9.1	40.7	4.5x
SPARTIAN	3.9	23.0	5.9x

+ tested on AMD MI210 GPU



# KOKKOS: GPU PORT OF A MACHINE-LEARNING APPLICATION

**SISSO++ (a machine-learning code based on compressed sensing / symbolic regression by the Fritz-Haber Institute of the MPG / NOMAD CoE)**

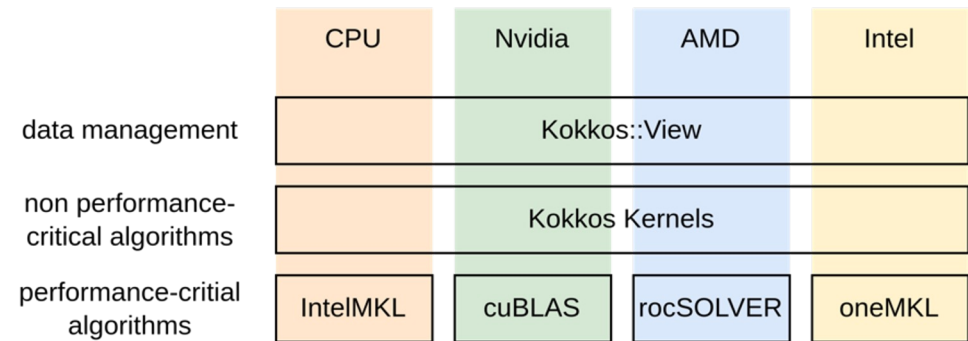


Kokkos porting based on:

- Kokkos views
- Kokkos lambdas

Hotspot: “batched” linear algebra (DGELS)

- hard to achieve with a general abstraction  
→ rely on vendor specific libraries
- provide nice encapsulation
- use batched BLAS to reduce launch overhead
- developed custom solver, outperforms cuBLAS



[https://gitlab.mpcdf.mpg.de/nomad-lab/cpp\\_sisso](https://gitlab.mpcdf.mpg.de/nomad-lab/cpp_sisso)



## MY OWN VIEW: *WHAT IF I HAD TO ...*

*... develop a new (GPU) code ?*

**use C++ (+ Kokkos, ...)**

*... port an existing CPU Fortran code to GPU ?*

**use OpenMP**

*... port an existing CUDA code to multiple GPU platforms ?*

**hippify for AMD GPUs (and see what happens with other GPU vendors)**

*... develop or port a CPU code or library with highest ambition for performance and longevity ?*

**use vendor-specific language (CUDA, HIP, SYCL) encapsulated by software-abstraction layer**

**watch out for consolidation opportunities (the SYCL promise, OpenMP, ...)**