

Make it so!

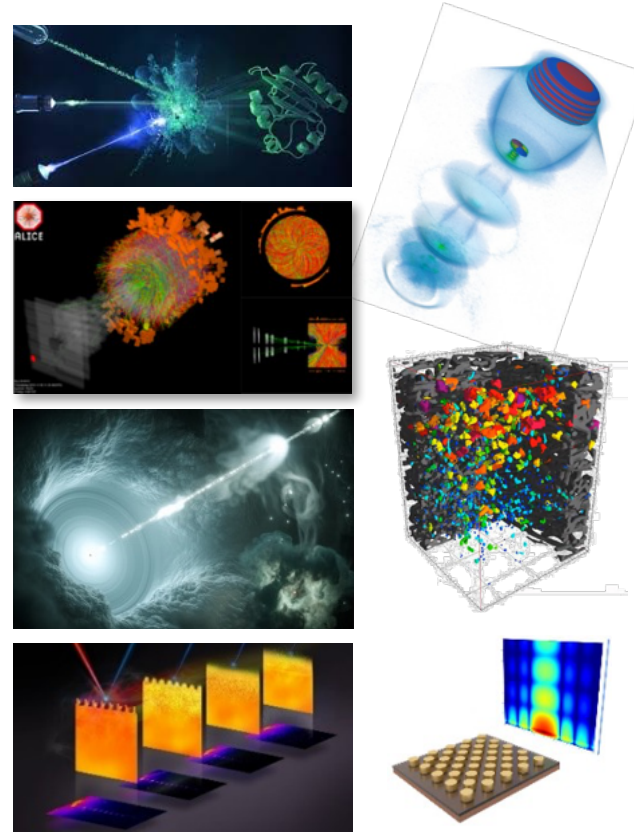
A fresh look on how individual researchers and teams can make large-scale research infrastructures work for them.



Helmholtz MATTER

Helmholtz MATTER

Large-scale research infrastructures for fundamental research



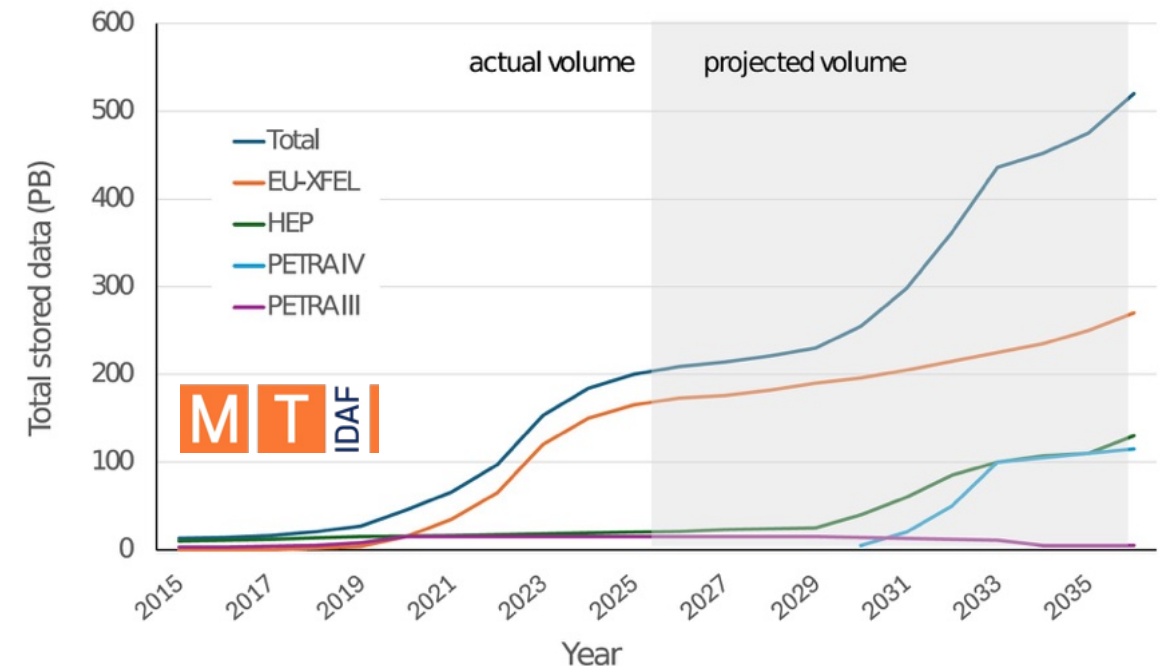
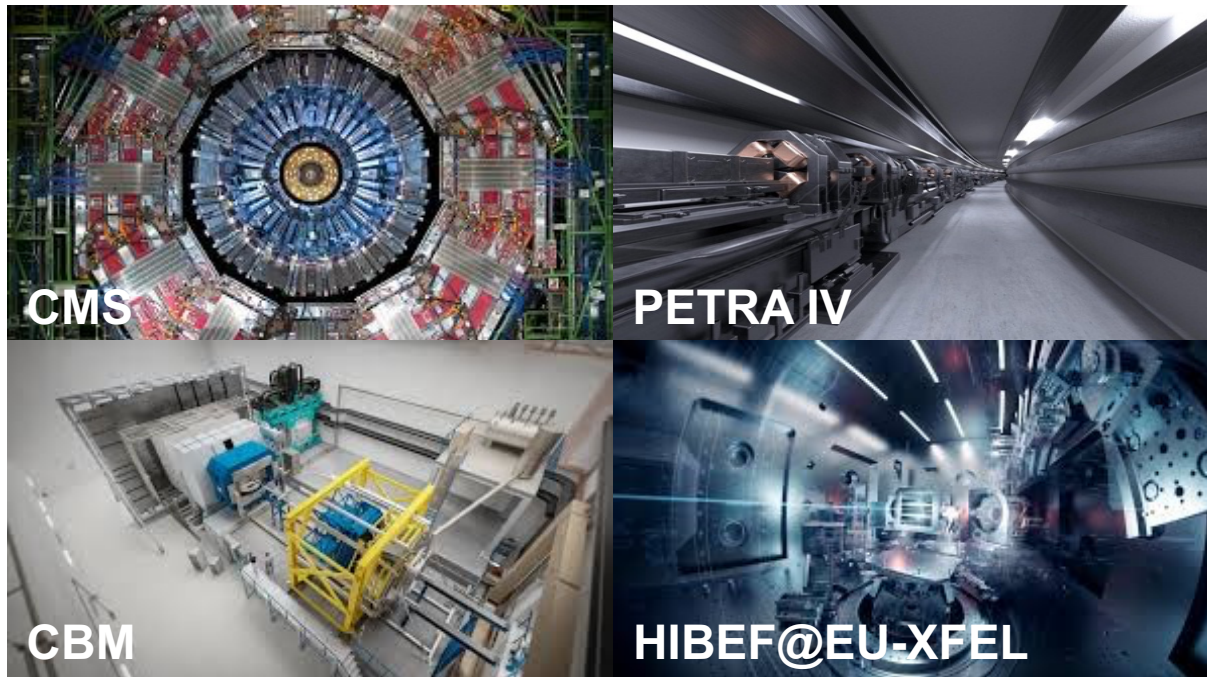
	LK II (User facilities)	International Research Infrastructures	National Research Institutes/ Infrastructures
DESY	FLASH PETRA III IDAF	LHC Belle II CTA (<i>under construction</i>) IceCube European XFEL ESRF	CFEL CSSB NanoLab DESY Test Beams DAF HIB@European XFEL PITZ
FZJ	JCNS (in MLZ)	ESS (<i>under construction</i>) ILL	(FRM-II)
GSI	UNILAC SIS18 ESR	FAIR (<i>under construction</i>) ALICE@LHC	HI Jena HI Mainz
Hereon	GEMS	ESS (<i>under construction</i>)	EMSC
HZB	BESSY II		SupraLab EMIL
HZDR	ELBE HLD IBC	European XFEL EMFL ESRF	HIB@European XFEL DRESDYN
KIT	GridKa	KATRIN Auger IceCube	ATP FLUTE TLK SR Beamlines

Particle Physics, Photon Science, Neutron Science, Astroparticle Physics, Accelerator Physics, Detector Physics, Quantum Physics, Matter under Extreme Conditions, Material science, Life sciences

We unlock the potential of Helmholtz MATTER's data

DMA is the focal point for computational and data science in MATTER

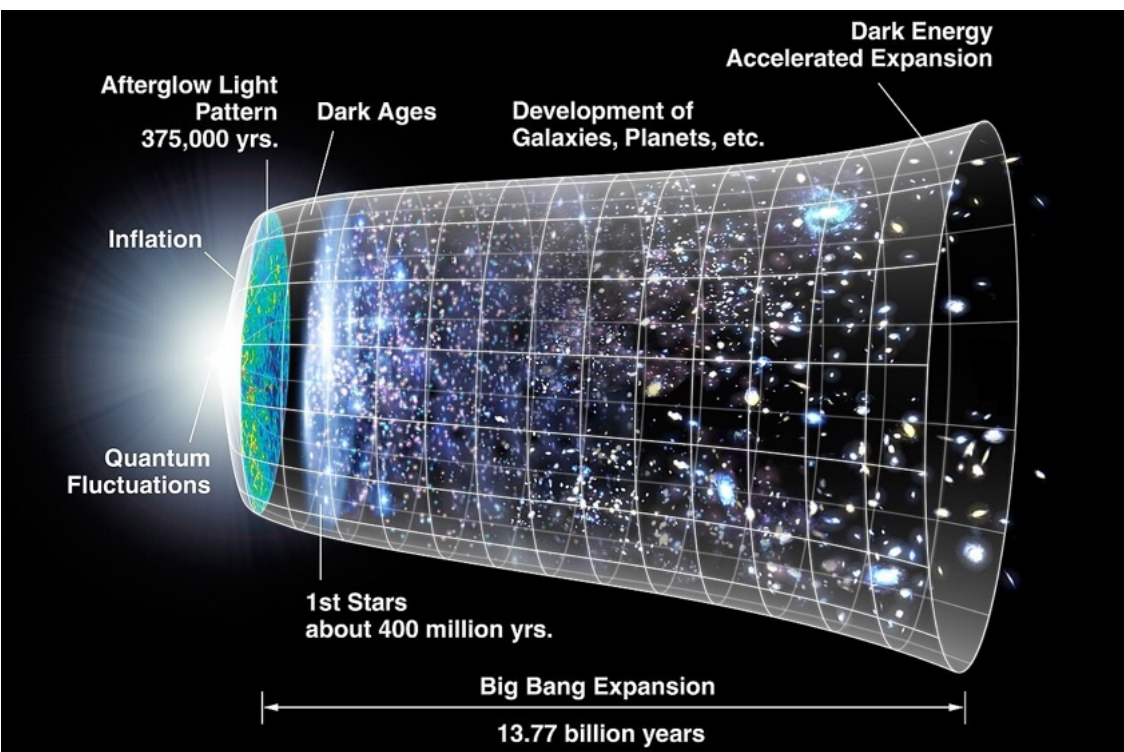
Helmholtz Matter research facilities produce **huge amounts of unique, high-quality data**. This is both a **challenge** and an **opportunity**.



Helmholtz manages & analyses data at high volumes and rates

We are world leading in LHC data management solutions and hold world records in data analysis

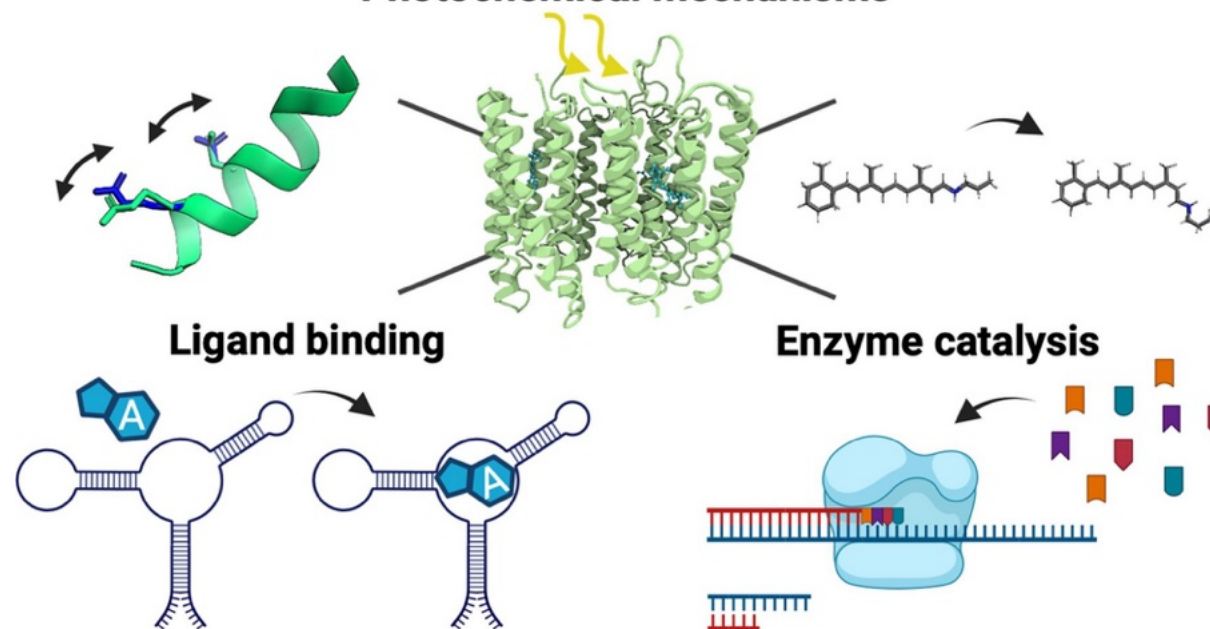
dCache **manages** >50% of remote LHC data



Crystfel **analyses** crystallograms at <500 ms world record speed

Time-resolved SX unravels atom-scale structural dynamics of:

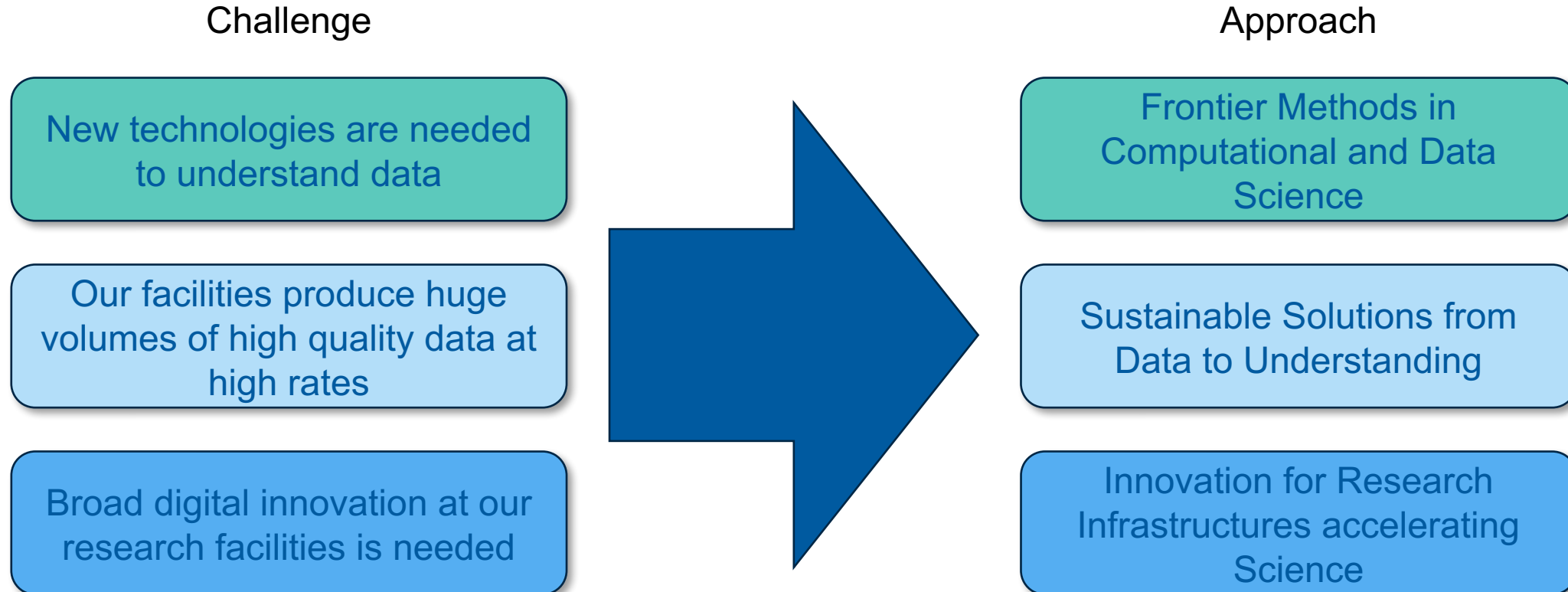
Photochemical mechanisms



Trends in Biochemical Sciences

Data Management and Analysis is a Helmholtz cross-center topic

Accelerating technological development transforms science and requires broad adaption



Autonomous Research Facilities

When does AI help?

Data + guided search/optimization + feedback (is there really nothing else?)

Lots of high quality data

Guided search / optimization

Feedback

How does AlphaFold2 work?

As part of AlphaFold2's development, the AI model has been trained on all the known amino acid sequences and determined protein structures.

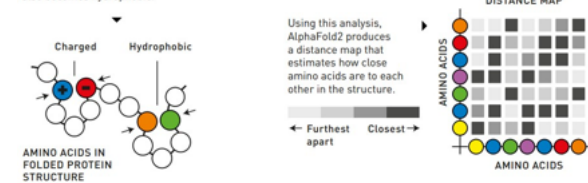
1. DATA ENTRY AND DATABASE SEARCHES

An amino acid sequence with unknown structure is fed into AlphaFold2, which searches databases for similar amino acid sequences and protein structures.

2. SEQUENCE ANALYSIS

The AI model aligns all the similar amino acid sequences – often from different species – and investigates which parts have been preserved during evolution.

In the next step, AlphaFold2 explores which amino acids could interact with each other in the three-dimensional protein structure. Interacting amino acids co-evolve. If one is charged, the other has the opposite charge, so they are attracted to each other. If one is replaced by a water-repellent (hydrophobic) amino acid, the other also becomes hydrophobic.

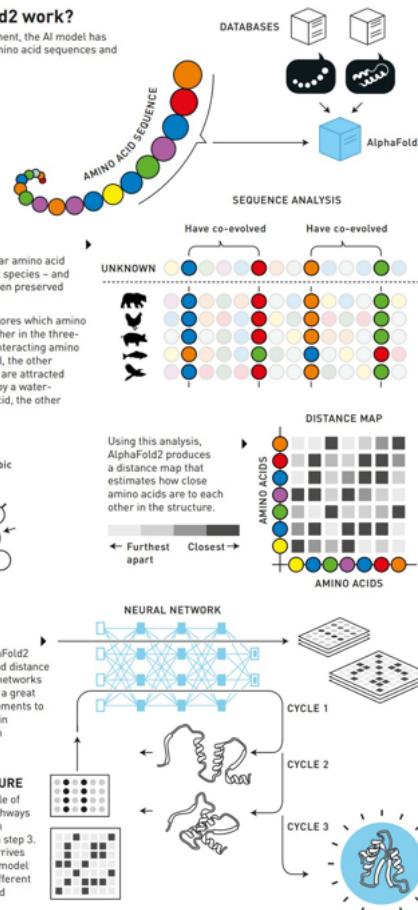


3. AI ANALYSIS

Using an iterative process, AlphaFold2 refines the sequence analysis and distance map. The AI model uses neural networks called transformers, which have a great capacity to identify important elements to focus on. Data about other protein structures – if they were found in step 1 – is also utilised.

4. HYPOTHETICAL STRUCTURE

AlphaFold2 puts together a puzzle of all the amino acids and tests pathways to produce a hypothetical protein structure. This is re-run through step 3. After three cycles, AlphaFold2 arrives at a particular structure. The AI model calculates the probability that different parts of this structure correspond to reality.



Planar Point Sets with Many Unit Distances

OpenAI

Abstract

For a finite planar set P , let $\nu(P)$ be the number of unordered unit-distance pairs in P , and let $\nu(n)$ be the maximum of $\nu(P)$ over all n -point planar sets. We prove that, for some fixed $\delta > 0$, one has $\nu(n) \geq n^{1+\delta}$ for infinitely many n . This disproves the well-known unit distance conjecture from [Erd46].

The construction passes through an infinite unramified tower of totally real number fields with 3-power Galois groups of growing degree, in which a fixed set of rational primes splits completely. After adjoining i , these fields produce high-dimensional lattices with many elements whose images under every complex embedding have absolute value 1. Golod-Shafarevich theory ensures existence of an infinite such tower, even after a quotient step making the prescribed Frobenius classes trivial. A crucial property of the construction is that all resulting discriminants and class numbers are at most exponential in the extension degree.

1 Main Result

For a finite set $P \subset \mathbb{R}^2$, let $\nu(P) = \#\{\{x, y\} \subset P : |x - y| = 1\}$ and $\nu(n) = \max_{|P|=n} \nu(P)$. The planar unit-distance problem goes back to Erdős [Erd46], who conjectured that there should be an absolute constant C such that, for all sufficiently large n ,

$$\nu(n) \leq n^{1+C/\log \log n}. \quad (1)$$

The same 1946 paper also introduced the distinct-distances problem, later resolved up to logarithmic factors by Guth–Katz [GK15]. An elementary upper bound is $\nu(n) = O(n^{3/2})$: the Euclidean unit-distance graph is $K_{2,3}$ -free, since two unit circles meet in at most two points, and the Kővári–Sós–Turán theorem applies [KST54]. The best known upper bound $\nu(n) = O(n^{4/3})$ is due to Spencer–Szemerédi–Trotter [SST84]; Székely later gave a short crossing-number proof of this incidence bound [Sze97].

A Tuning Prompt

Human: Now you will help me optimise a function with five input variables. Q1, Q2, Q3, Q4 and Q5. I have some Q1, Q2, Q3, Q4, Q5 pairs and the corresponding function values at these points. The pairs are arranged in descending order based on their function values, where lower values are better.

Inputs:

```
Q1: -13.30,
Q2: -4.65,
Q3: -3.00,
Q4: -4.00,
Q5: -6.50
```

Objective value = 2.37

Task description

Below are previously measured pairs of magnet settings and the corresponding observed beam parameters.

Magnet settings:

```
Q1: -13.30,
Q2: -4.65,
Q3: -3.00,
Q4: -4.00,
Q5: -6.50
```

Beam parameters:

```
Q1: -13.30,
Q2: -4.65,
Q3: -3.00,
Q4: -4.00,
Q5: -6.50
```

Give me new magnet settings that are different from all pairs above. The magnet settings you should propose should lead to beam parameters closer to the target. If you do not have any magnet settings to propose, please return the last magnet settings as proposed.

The output should be a markdown code snippet formatted in the following schema, including the leading and trailing "```" and "```".

Previous samples

Output instructions

B Optimization Prompt

Human: Now you will help me optimise a function with five input variables. Q1, Q2, Q3, Q4 and Q5. I have some Q1, Q2, Q3, Q4, Q5 pairs and the corresponding function values at these points. The pairs are arranged in descending order based on their function values, where lower values are better.

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Previous samples

Output instructions

C Chain-of-Thought Prompt (based on Explained Prompt)

Human: Now you will help me optimise a function with five input variables. Q1, Q2, Q3, Q4 and Q5. I have some Q1, Q2, Q3, Q4, Q5 pairs and the corresponding function values at these points. The pairs are arranged in descending order based on their function values, where lower values are better.

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```

Beam parameters:

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Previous samples

Output instructions

D Explained Prompt (based on Tuning Prompt)

Q1, Q2 and Q3 are quadrupole magnets. When their k1 strength is increased, the beam becomes more focused in the horizontal plane and more defocused in the vertical plane. When their k1 strength is decreased, the beam becomes more defocused in the horizontal plane and more focused in the vertical plane. When their k1 strength is zero, the beam is not focused in either plane. Quadrupole magnets might also affect the beam in the longitudinal or vertical plane depending on their Q2 deflection, where the beam does not pass through the centre of the magnet. The range of the k1 strength is -30.0 to 30.0 mT/m.

Q4 is vertical steering magnet. When its deflection angle is increased, the beam is steered upwards, where its deflection angle is decreased, the beam is steered downwards. The range of the deflection angle is -4.0 to 4.0 mrad.

Q5 is horizontal steering magnet. When its deflection angle is increased, the beam is steered to the right, where its deflection angle is decreased, the beam is steered to the left. The range of the deflection angle is -4.0 to 4.0 mrad.

Previous samples

Output instructions

E Response

Human: Now you will help me optimise a function with five input variables. Q1, Q2, Q3, Q4 and Q5. I have some Q1, Q2, Q3, Q4, Q5 pairs and the corresponding function values at these points. The pairs are arranged in descending order based on their function values, where lower values are better.

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Q1: -13.30,
Q2: -4.65,
Q3: -3.00,
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```

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Task description

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Q5: -6.50
```

Beam parameters:

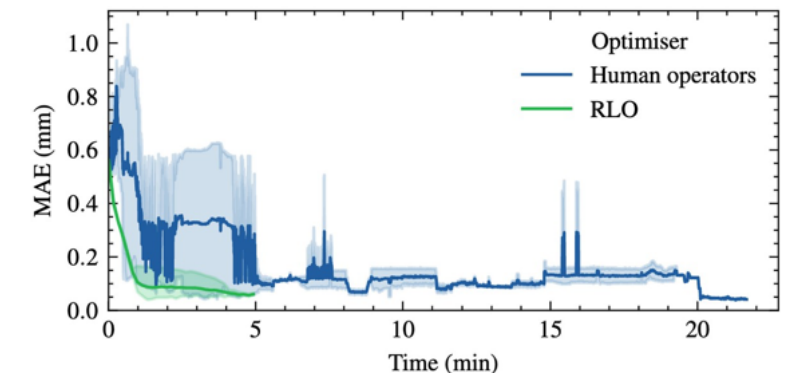
```
Q1: -13.30,
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Q3: -3.00,
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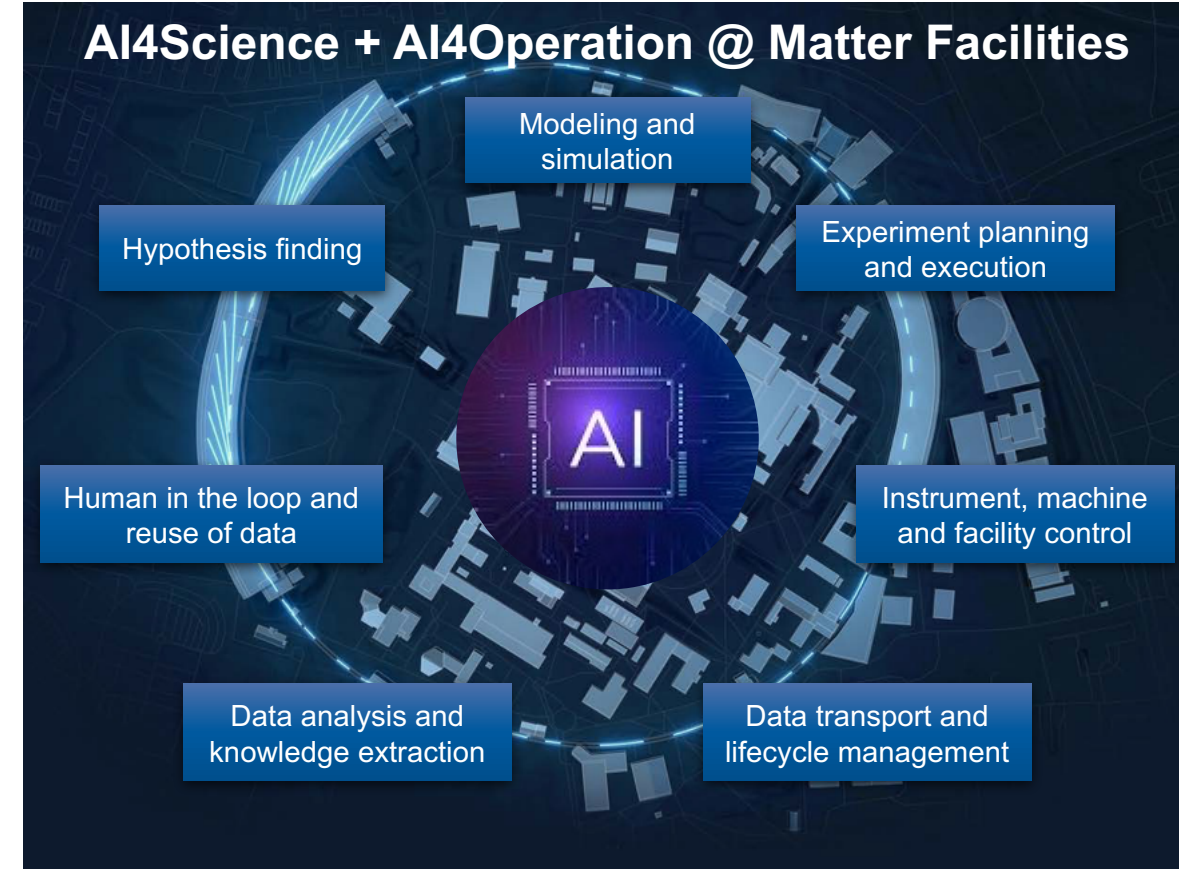
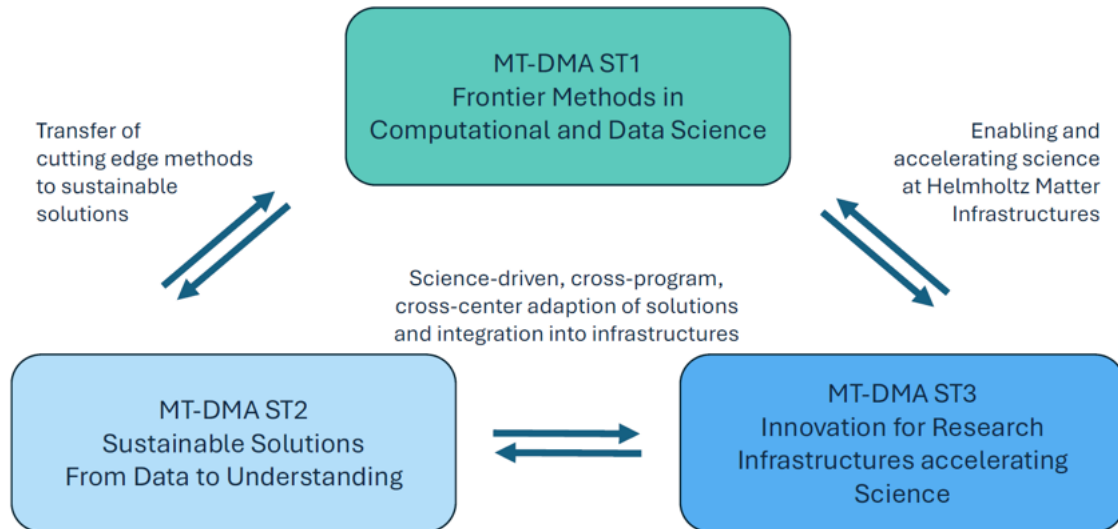
Previous samples

Output instructions



Helmholtz common vision of autonomous facility operation

A shared approach across research centers and disciplines



Autonomous Research Facilities

World models include the machines, detectors, models, scientists, ...

- **A well defined understanding of observables**
 - A clear ontology
 - Models with predictive capability
 - How to measure and quantify observables
- **A complete knowledge of your system**
 - Digital twins of systems under investigation
 - Variation over all possible outcomes
- **Digitalization of facility operation and experiments**
 - A quantifiable and steerable model of your facility
 - Include all known sources of errors
 - A complete error analysis
 - Invertible synthetic diagnostics
- **Useable and useful F.A.I.R. workflows**
 - for simulation
 - for data analysis
 - for visualization
 - for operation and experiments

Autonomous Research Facilities

World models include the machines, detectors, models, scientists, ...



Dagstuhl Seminar 23132 / Dagstuhl Reports 13
AI-Augmented Facilities:
Bridging Experiment and Simulation with ML



HPC+AI 4 Fusion

Frontier computing for fusion science

Joining forces in Europe, bringing together magnetic and laser-driven fusion

EuroHPC Center of Excellence for Plasma Simulations PLASMA-PEPSC



EuroHPC
Joint Undertaking



- Brings **laser fusion** and **magnetic fusion** community together
- Common modeling efforts for atomic physics, looking at **plasma-wall interaction** (magnetic fusion) and **radiation transport** (laser fusion)

How to run on every TOP 500 System in record time

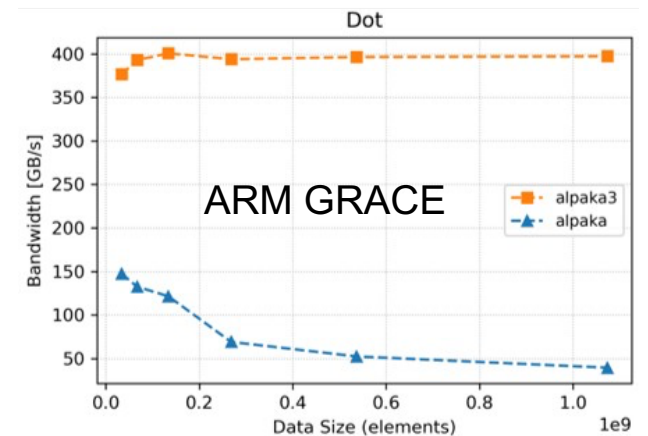
Performance-portable parallel programming on heterogeneous systems



Alpaka: The performance portable C++ abstraction library for HPC

- “The European Kokkos”
- SIMD + SIMT
- Write code one time and run it everywhere
- Supports GPU (Nvidia, AMD, Intel) and CPU (x86, ARM, RISC-V)
- Minimum overhead
- Follows the default GPU offload model
- Single source C++20 user friendly API
- Allows to use native vendor APIs like CUDA, HIP and OneAPI

GH 200



Large-scale simulations for fusion research

Making the best of the worlds fastest supercomputers



Institute

Jülich Supercomputing Centre (JSC)

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Research

Systems

Services

Education

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🏠 / JUPITER - Exascale for Europe / Applications: Make JUPITER shine

Applications: Make JUPITER shine

Applications are a central aspect in the [exascale strategy](#) of the Jülich Supercomputing Centre (JSC). Concrete use cases computed on JUPITER have the potential to advance developments in important areas of society and, in particular, to significantly accelerate change due to JUPITER's enormous total computing power. JUPITER will achieve more than 1 ExaFLOP/s for double-precision computing operations and more than 70 ExaFLOP/s for 8-bit precision, relevant for AI applications in particular.

JUPITER Benchmark Suite

Applications played an important role in the JUPITER procurement process. The JUPITER Benchmark Suite was developed to bring a system into production that not only performs very well in synthetic benchmarks but also in real and relevant use cases. Based on current and projected important applications, a collection of 16 applications was compiled and used for the performance evaluation of possible JUPITER configurations. Andreas Herten, lead author of the corresponding [scientific paper](#), comments: "With the JUPITER Benchmark Suite, we have created a milestone for the a-priori evaluation of supercomputers. The combination of standardization and real applications is unique and allowed a targeted projection of the expected performance."

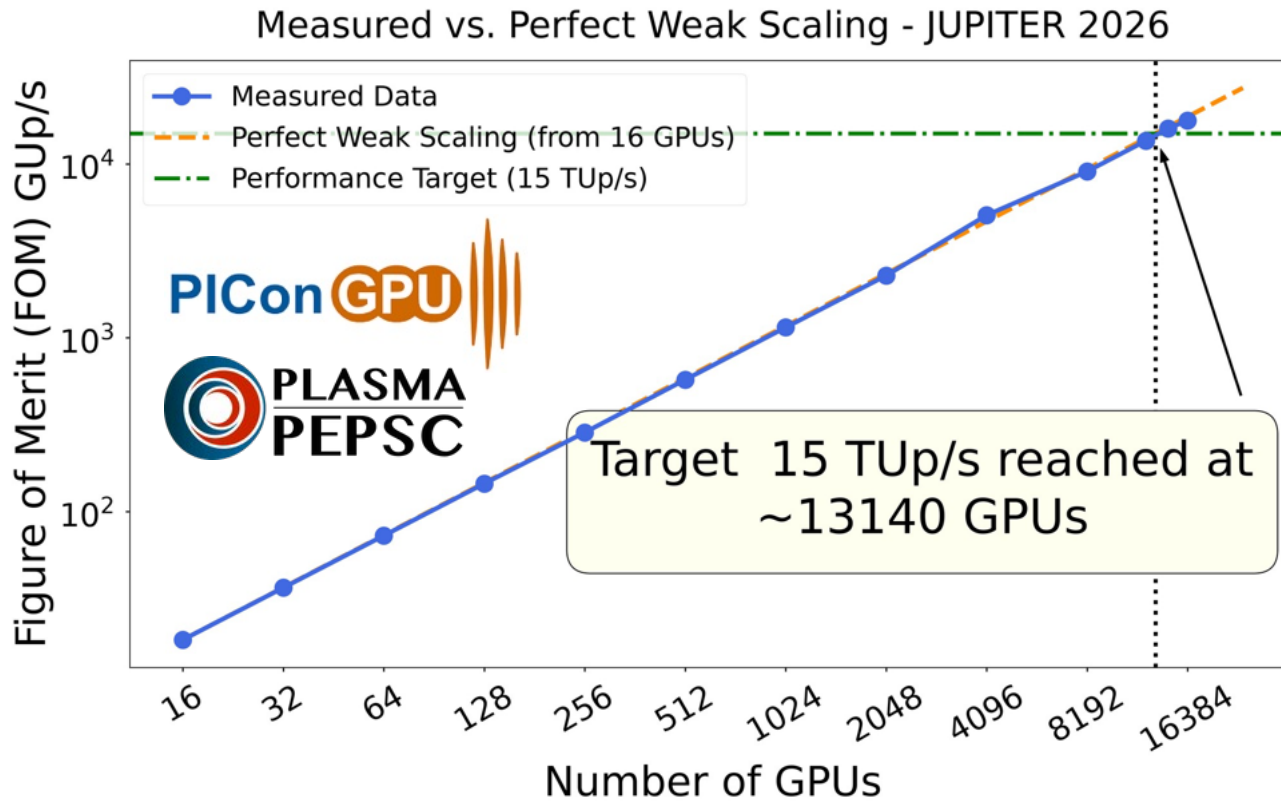
JUPITER Research and Early Access Program

For applications, JUPITER is not only an opportunity but also a challenge. To use a supercomputer of such size to its full potential, applications must be thoroughly prepared and optimized. This can range from making adjustments to algorithms and methods to optimizing individual settings. Such optimizations are currently being carried out as part of the JUPITER Research and Early Access Program (JUREAP). In JUREAP, experts from JSC are working with more than 100 applications to assess and prepare them for exascale. The applications are highly diverse and have the potential to fundamentally advance many different scientific disciplines. They range from AI applications, such as foundation models and generative video generation, to climate models, particle physics and energy applications, to molecular dynamics simulations relevant for drug development and disease control. Some examples are briefly presented below.

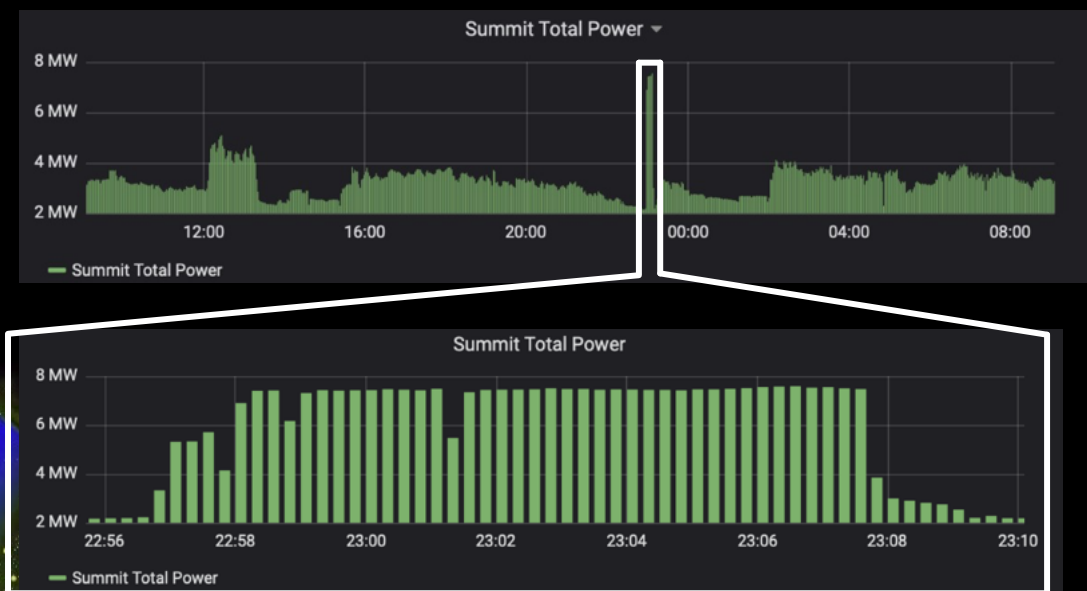


Large-scale simulations for fusion research

Making the best of the worlds fastest supercomputers

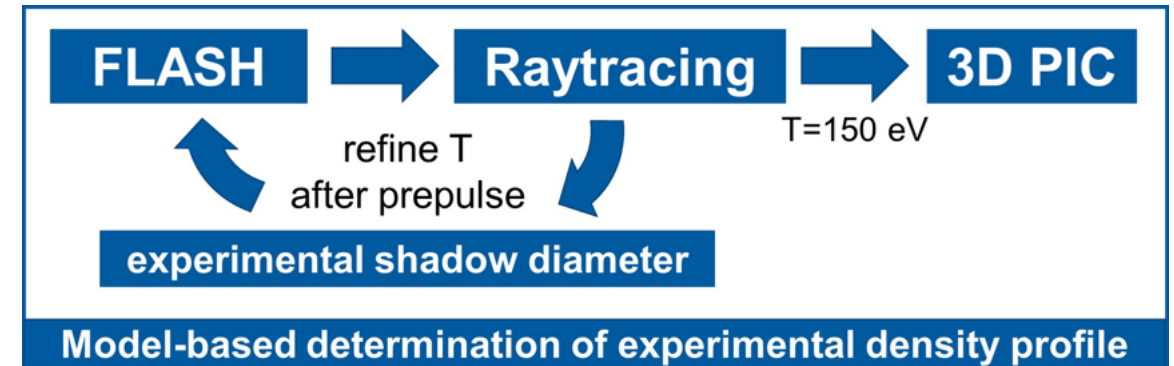
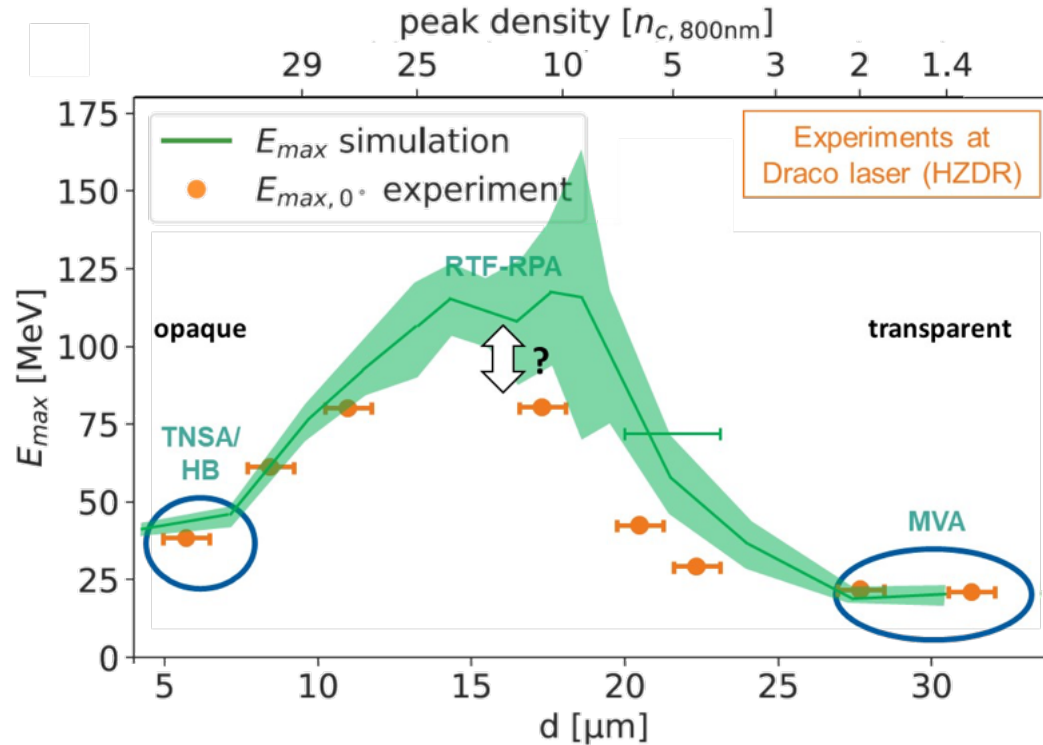


High energy density physics
@ fs & nm resolution
@ PB/s data rates
@ 20 fps
@ o(10) MW power consumption



World models combining experiment and simulation

A well defined understanding of observables: Simulate what is measured

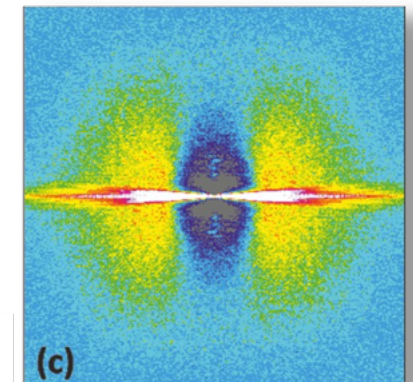
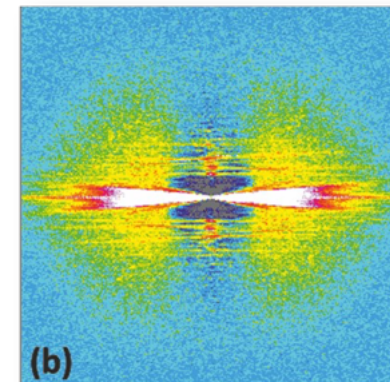
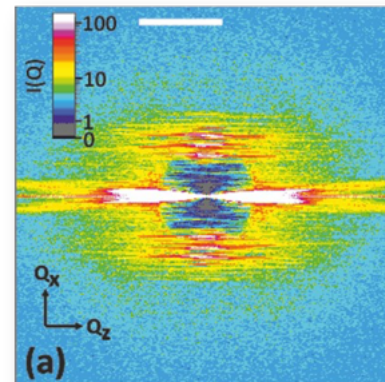
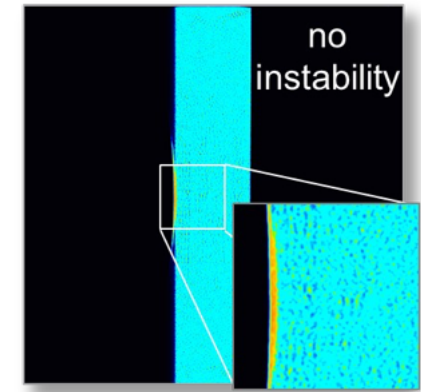
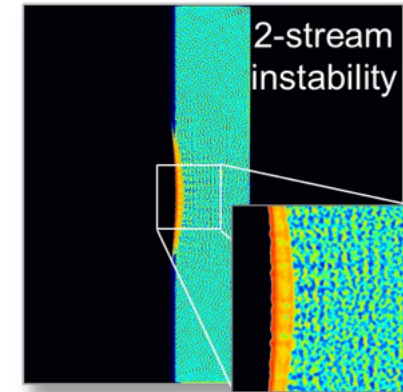
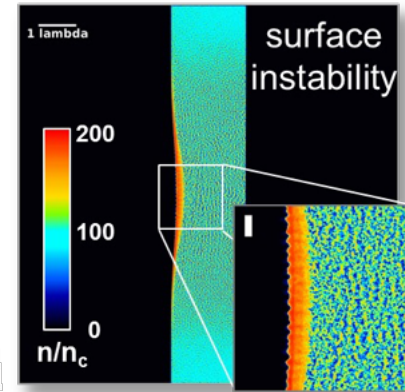
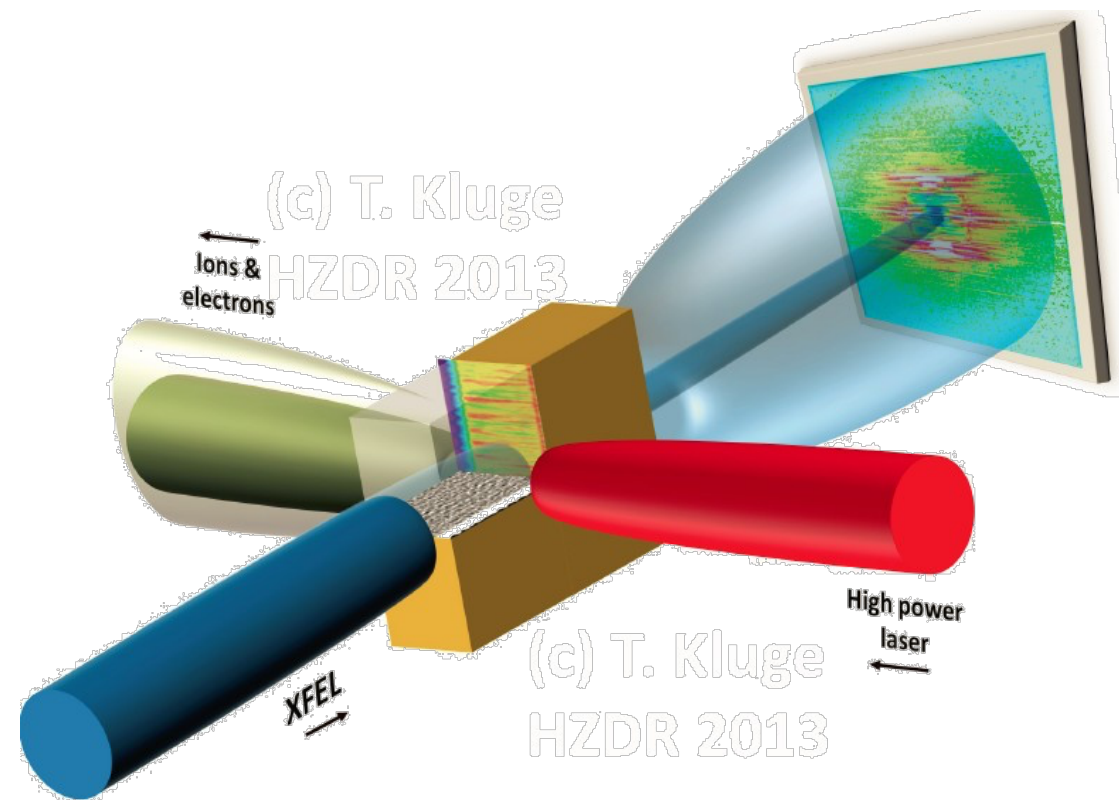


T. Kluge + K. Zeil and many others

HiBEF HELMHOLTZ INTERNATIONAL
BEAMLINE FOR EXTREME FIELDS

World models combining experiment and simulation

A complete knowledge of your system: Atomic resolution of high energy density matter

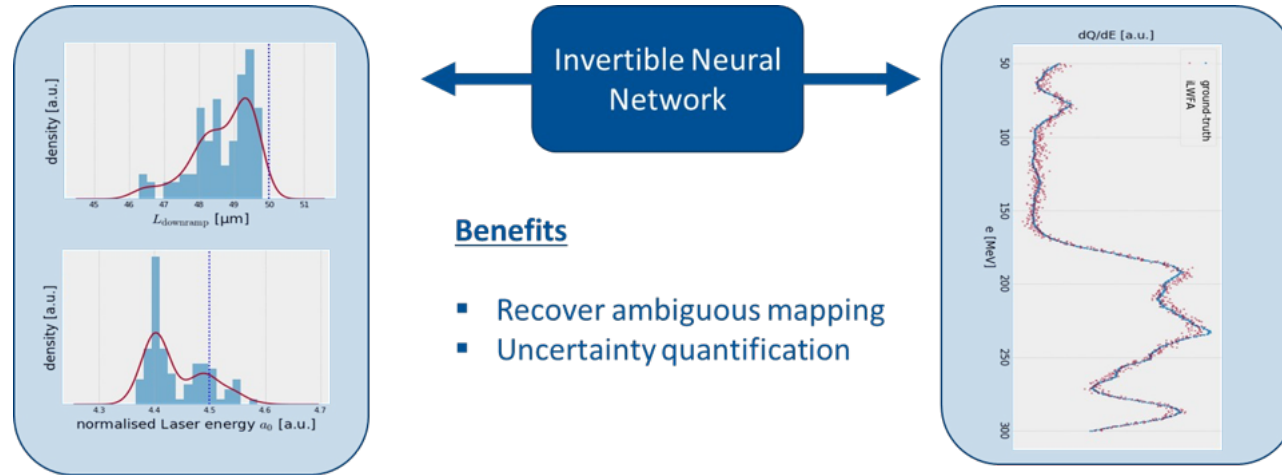


T. Kluge, L. Huang

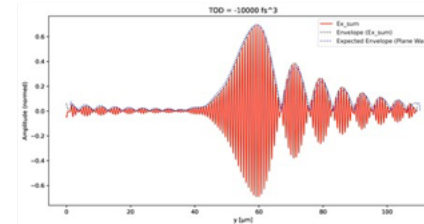
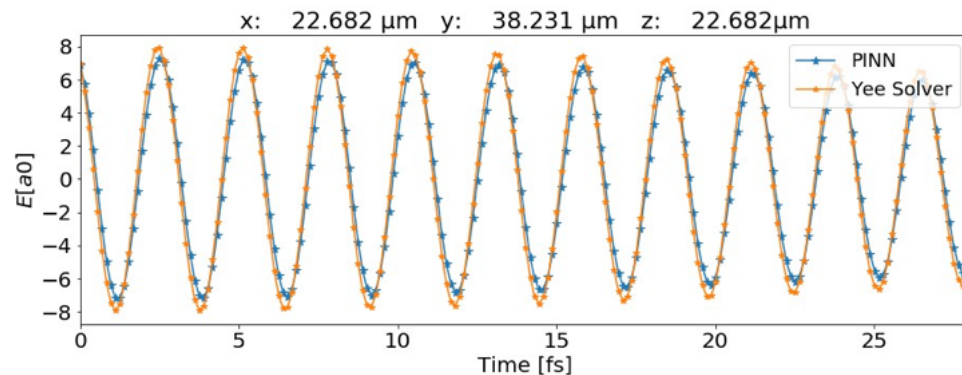
World models combining experiment and simulation

Digitalization of facility operation and experiments: Invertible models for discovery

Invertible physics-informed neural networks

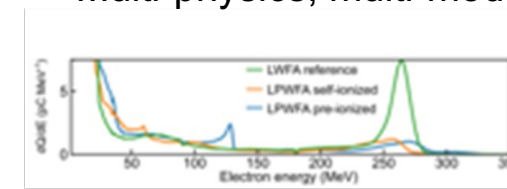


ML-based solvers

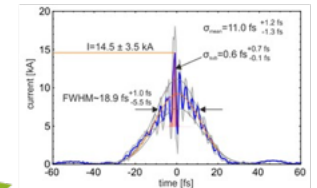


Experimental parameters

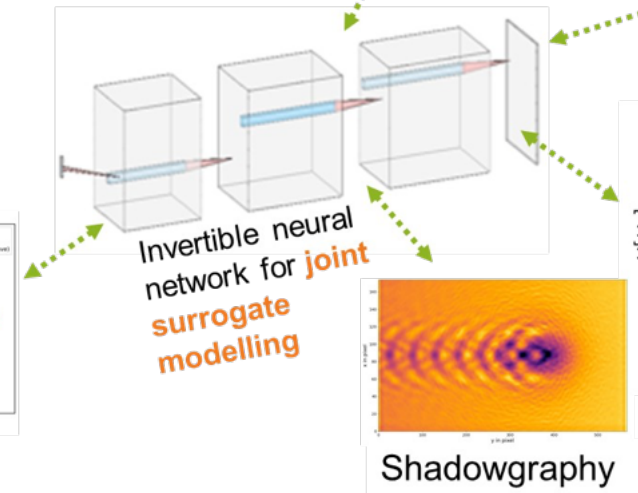
Multi-physics, multi-modal surrogate modeling



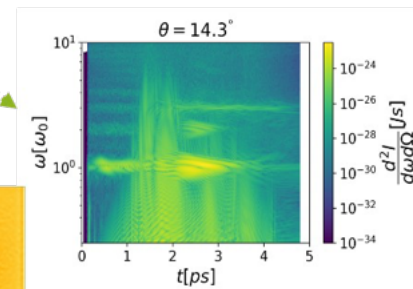
Particle energy spectra



Transition radiation



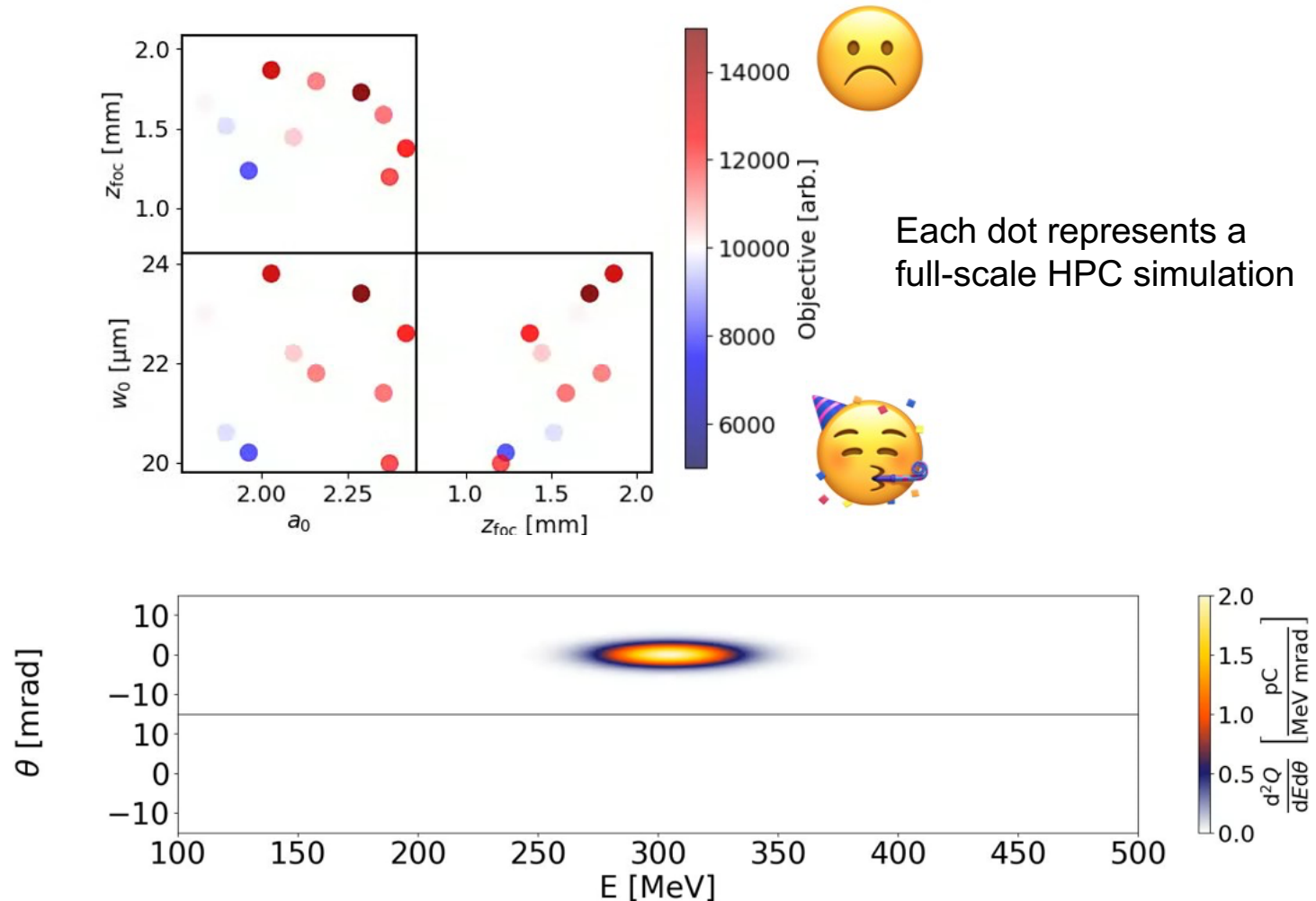
Shadowgraphy



Radiation spectra

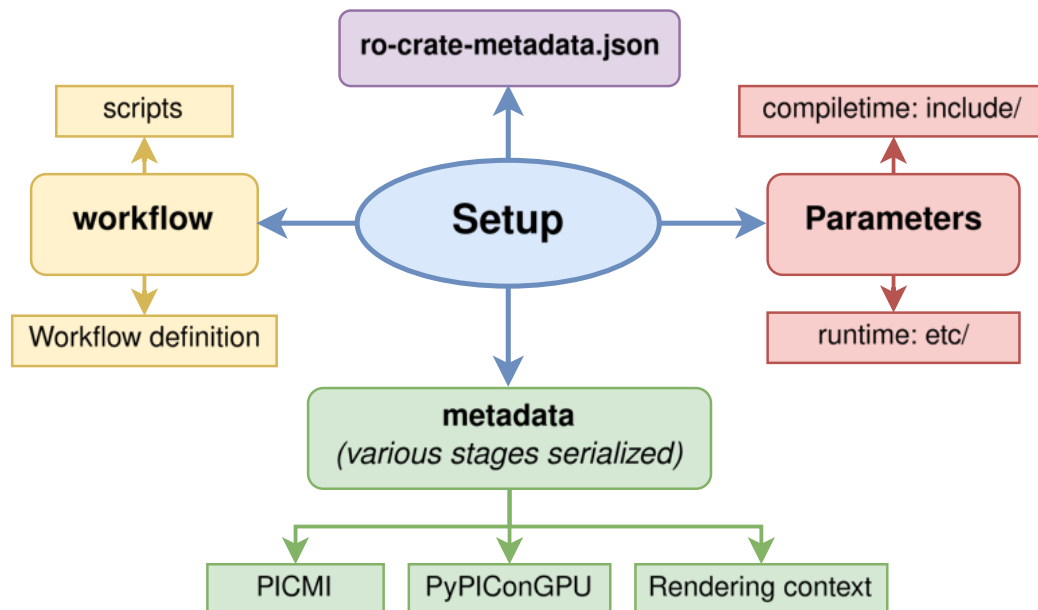
World models combining experiment and simulation

Digitalization of facility operation and experiments: Campaign-driven optimized operation



World models combining experiment and simulation

Useful and useable workflows: Community-driven input/output standards for workflows



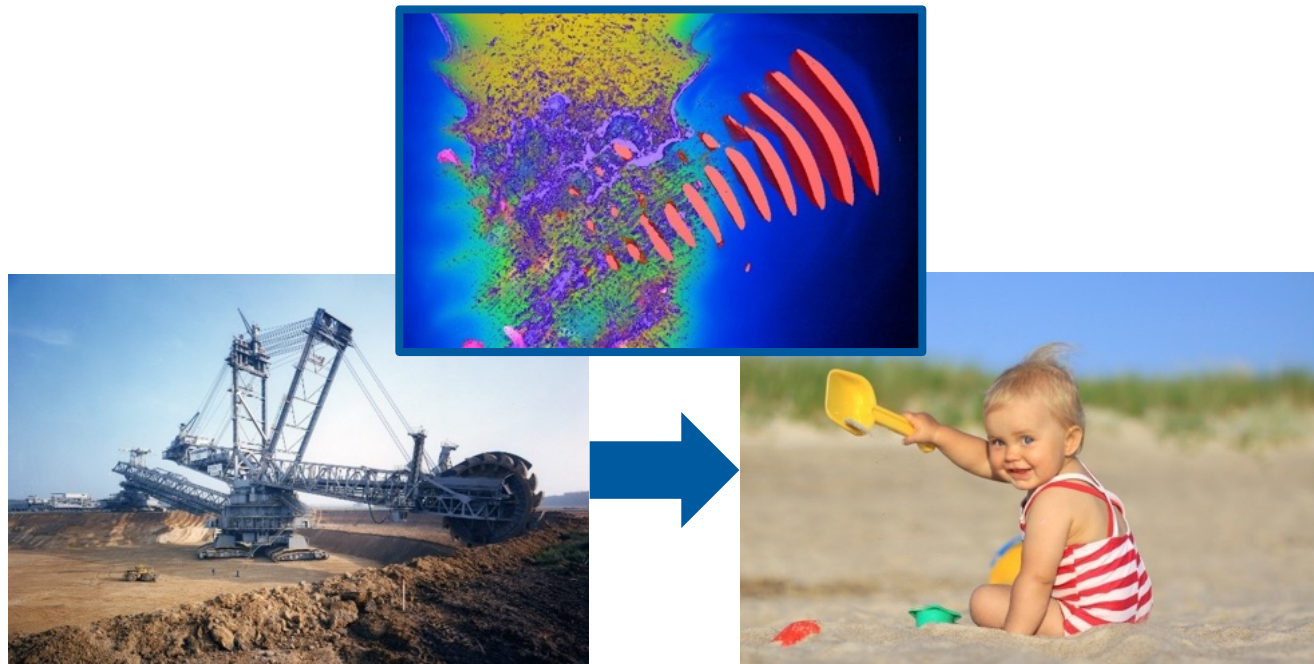
```
from picongpu import picmi

sim = picmi.Simulation(
    time_step_size=9.65531e-14,
    max_steps=1024,
    solver=picmi.ElectromagneticSolver(
        method="Yee",
        grid=picmi.Cartesian3DGrid(...)
    ),
    ...
)

sim.run()
```

Yet, data storage is a thing of the past!

Data rates become too high, volumes too large for storage



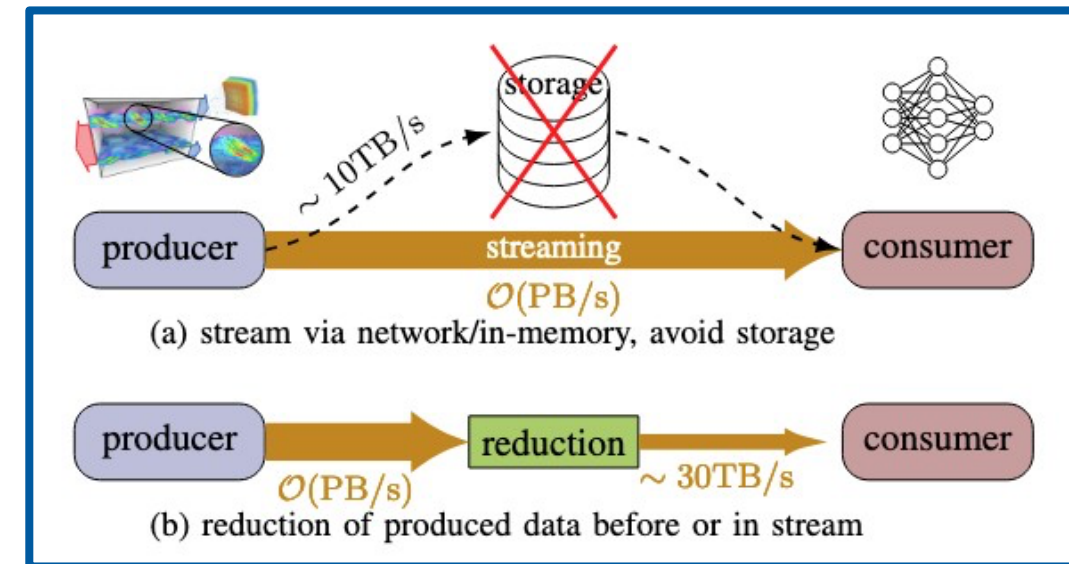
0.6 PetaByte / s

0.1TB/s GPU to GPU 16GB/s internode

file System: 679 PB

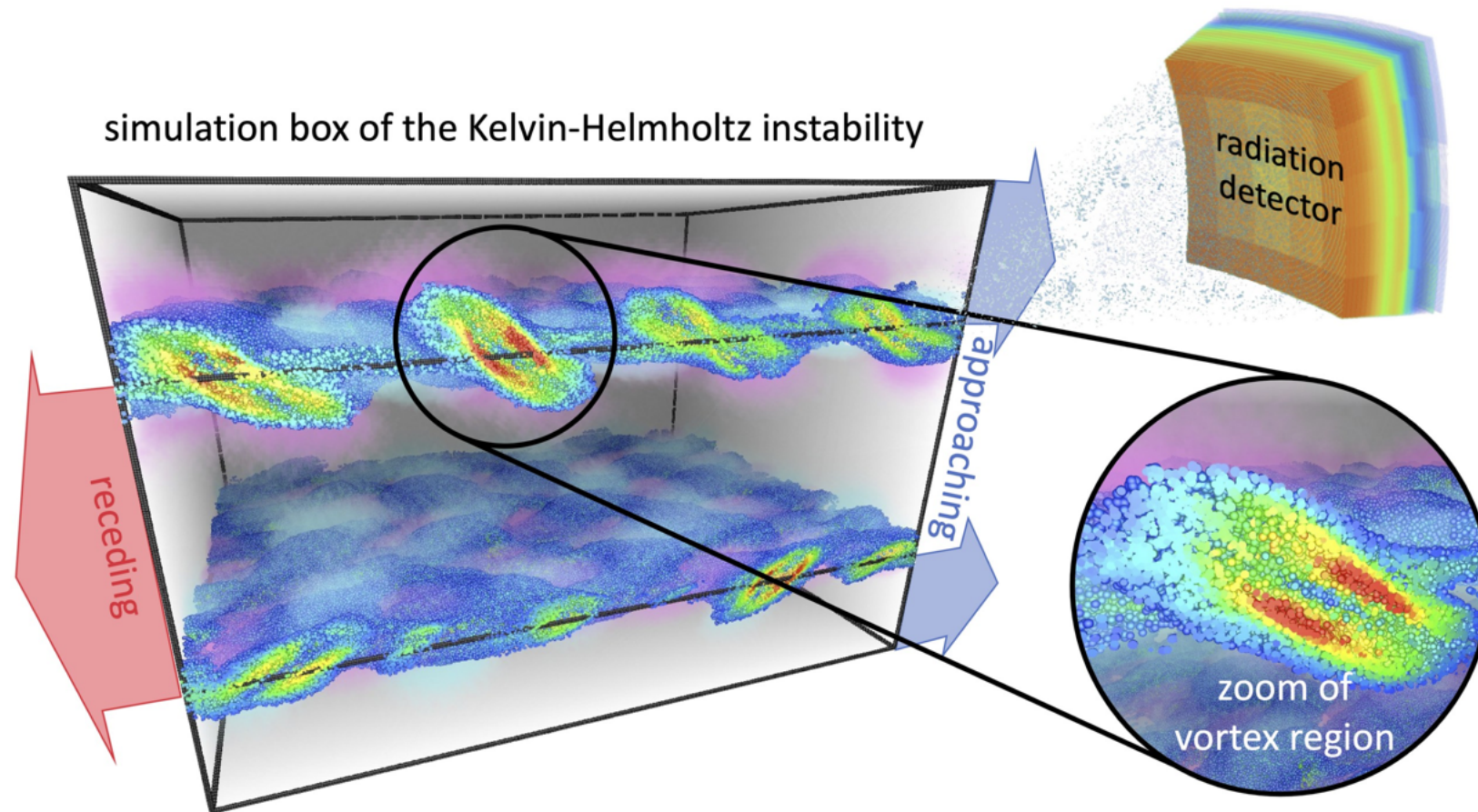
full in 100s

but simulation runs longer



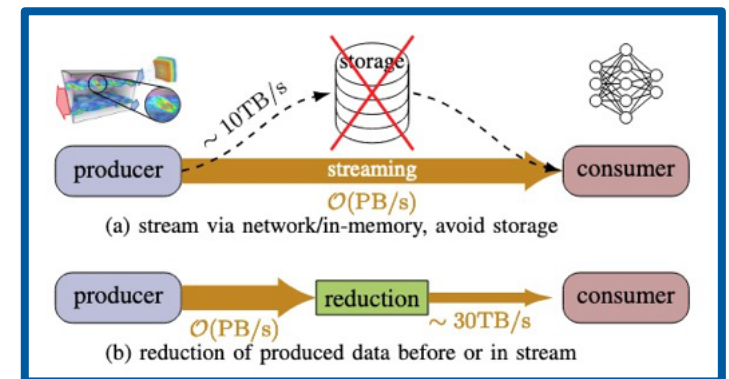
Coupling HPC with AI training at Exascale

Training of a large-scale invertible AI model with streaming simulation data



Train neural networks for:

- Forward problem:
particle to radiation
- Compression:
particle to particle latent space
- Inverse problem:
radiation to particle

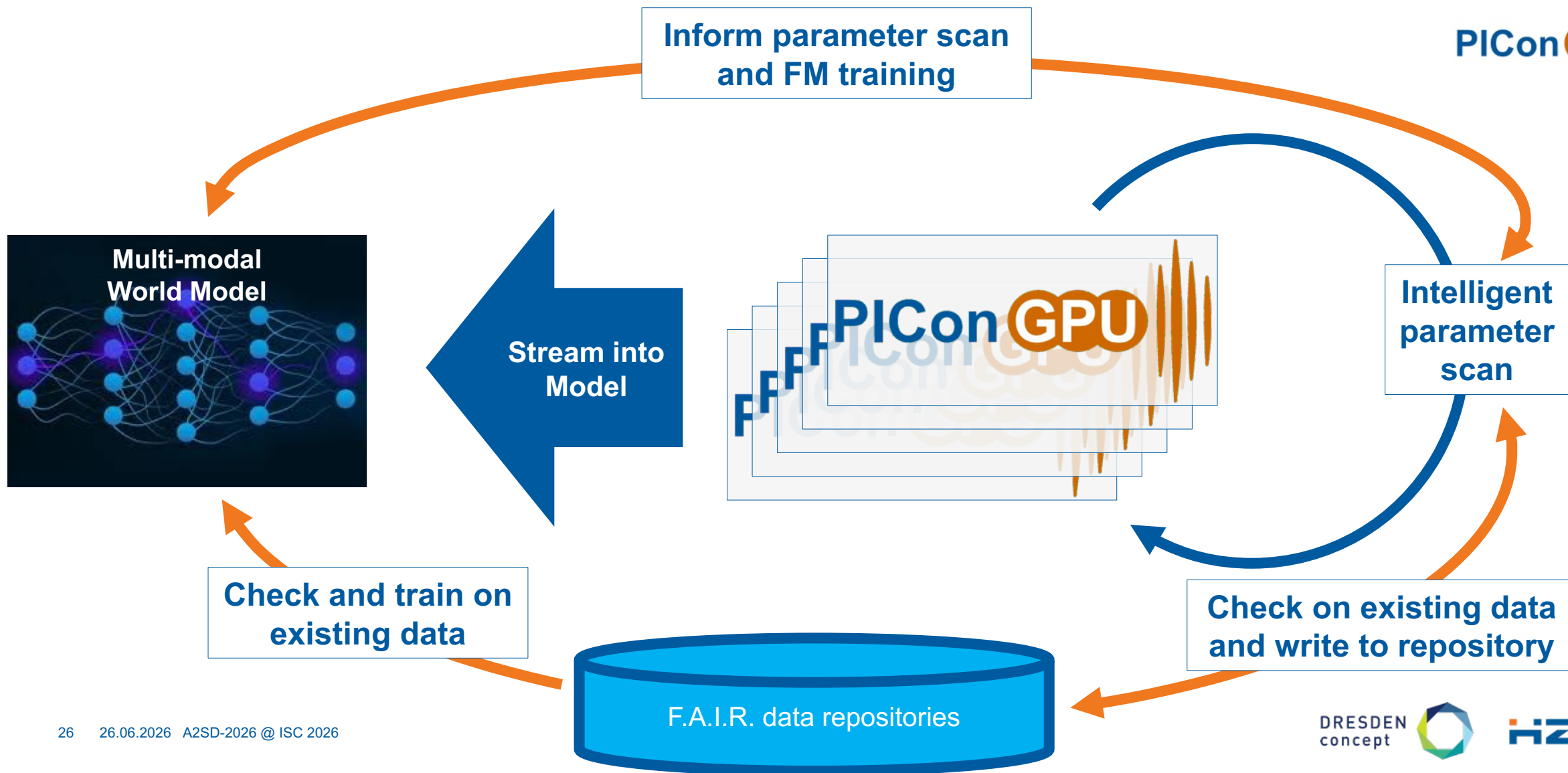


Online training of streaming data @ Exascale

Why use files when you have neural networks?

open
PMD

PICon GPU



Online training of streaming data @ Exascale

Why use files when you have neural networks?

open
PMD

PICon GPU

Inform parameter scan
and FM training

Intelligent
parameter
scan

Check on existing data
and write to repository

F.A.I.R. data repositories

Check and train on
existing data

Stream into
Model

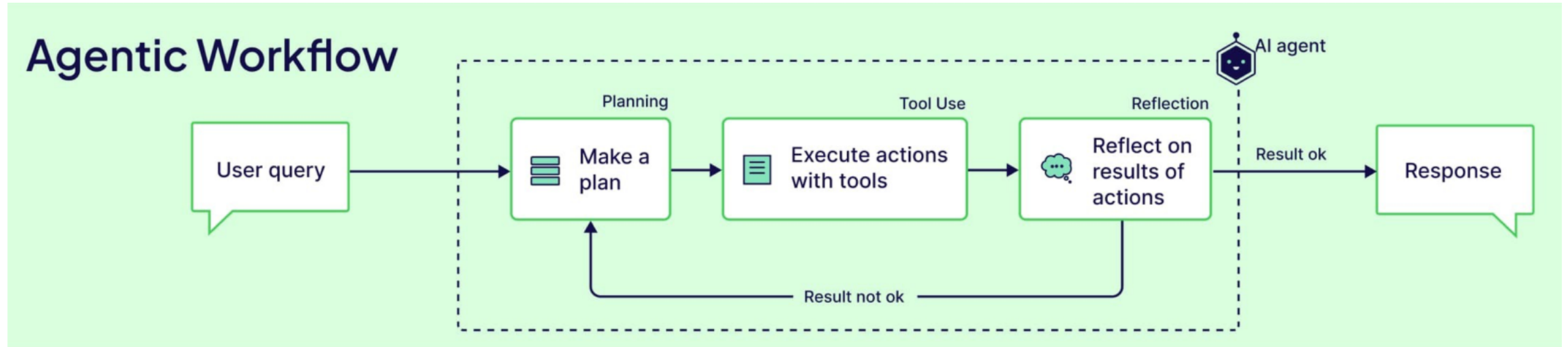
Multi-modal
World Model



Make it so!

Agentic AI puts the user in the center, not the data

Scientists know stuff and like to tinker



Agentic AI solves:

- Diverse user needs
- Easy usage

Agentic AI helps with:

- Tooling
- Interfaces
- Interoperability

Agentic AI is (currently) limited in:

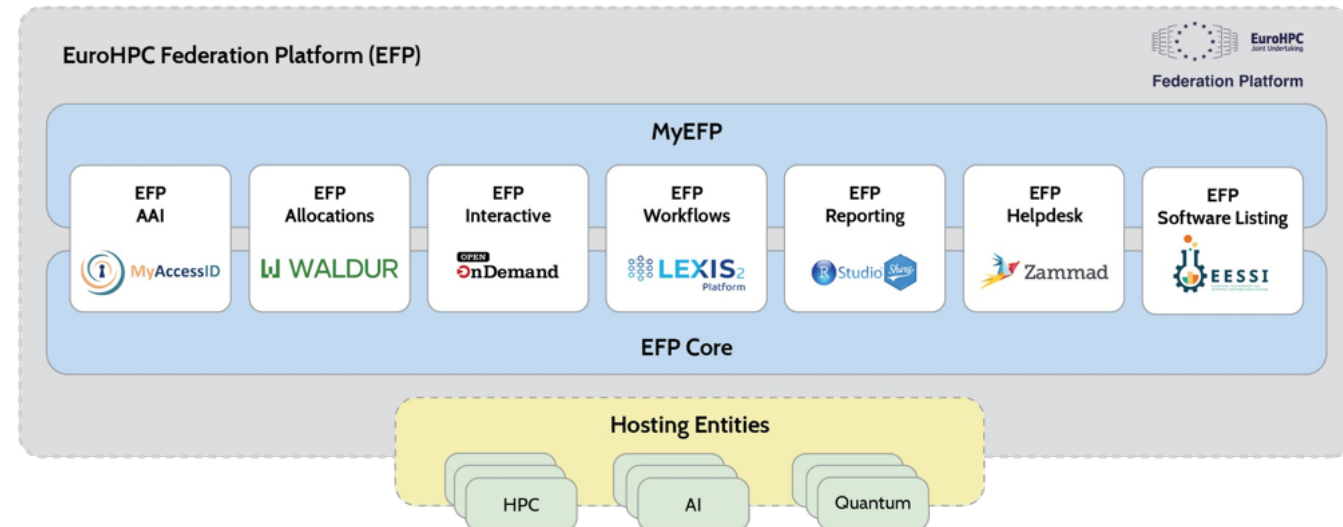
- Autonomous discovery*
- Optimization**

* Our research facilities are not (yet) flexible enough for agentic planning (all hail our robot overlords!)

** In many cases, an optimum operation point stays an optimum operation point

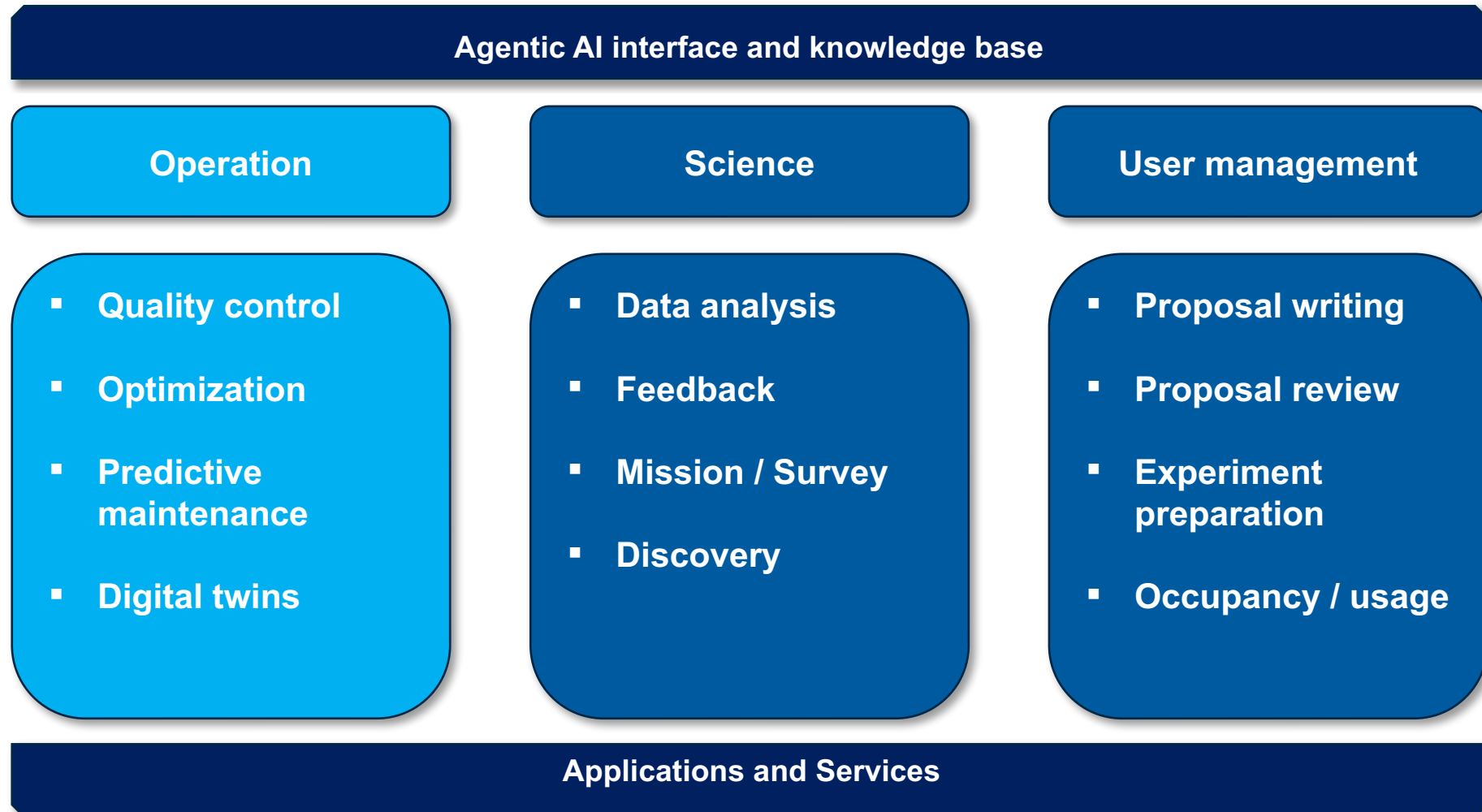
Stop selling shells on the seashore... sorry, in HPC!

Once we have clear interfaces, HPC clusters are just tools queried by agents



The three facility pillars: Operation, Science, User management

In Europe, there is a TIER-Systems of Compute/Data Centers, EuroHPC is still orthogonal to that



Make it so!

Autonomous user facilities as tools for scientists

Learnings:

- Analysis to Data
- Federation creates trust
- Include user communities
- Start with usage, not data
- Make data intake automatic
- Make it safe & useful

