

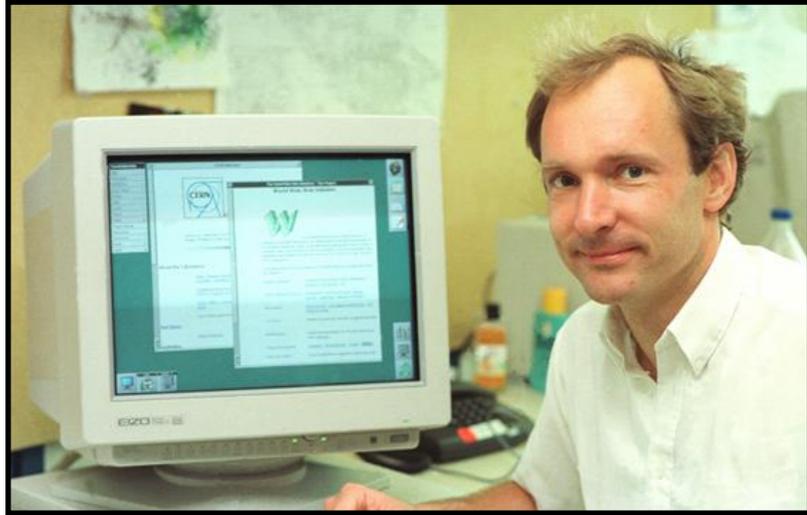
Input on AI

Gregor Kasieczka, Stefano Giagu

ESPP Meeting, Venice
June 2025

Disclaimer: Thankful for input by everyone. Bias and opinions in this synthesis are my own.

Development



Past

*WorldWideWeb:
Proposal for a HyperText
Project, Tim-Berners
Lee et al* **1990**



Present

Explosion of large-
language models
(LLMs) **2025**



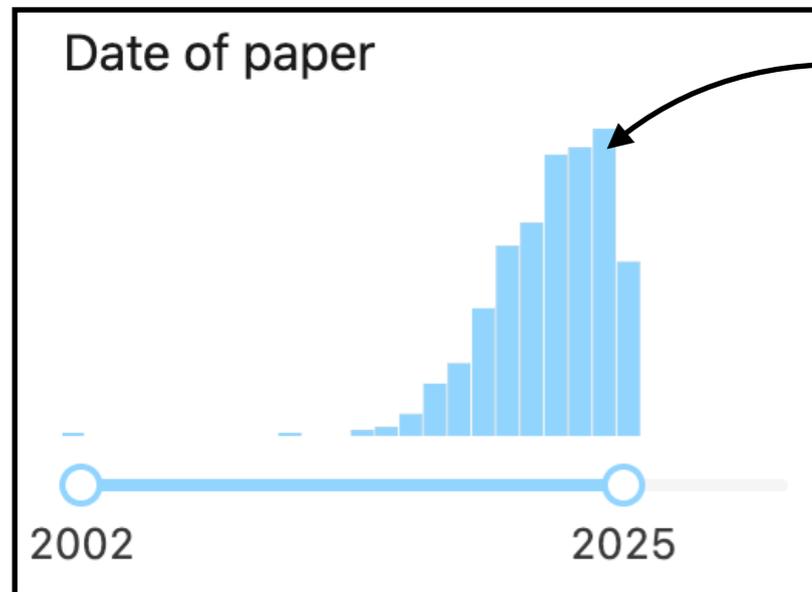
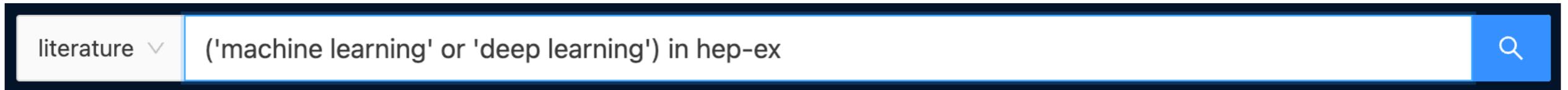
Future

???, **2060**

Cannot be too **ambitious** for the physics
possibilities from **developments in AI**

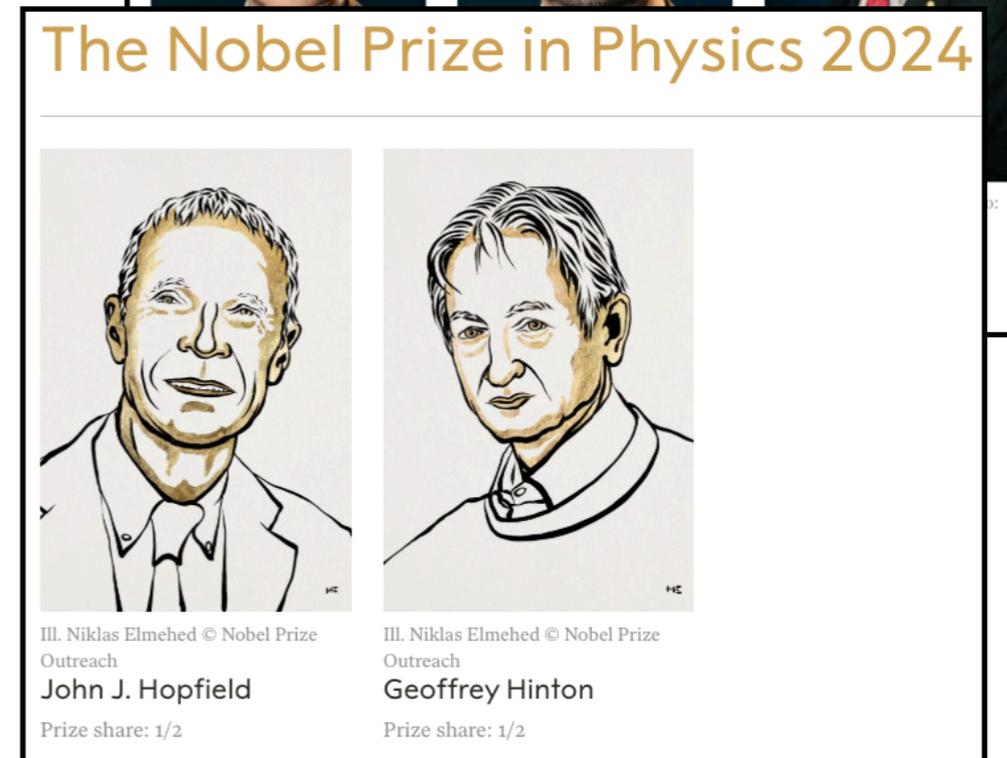
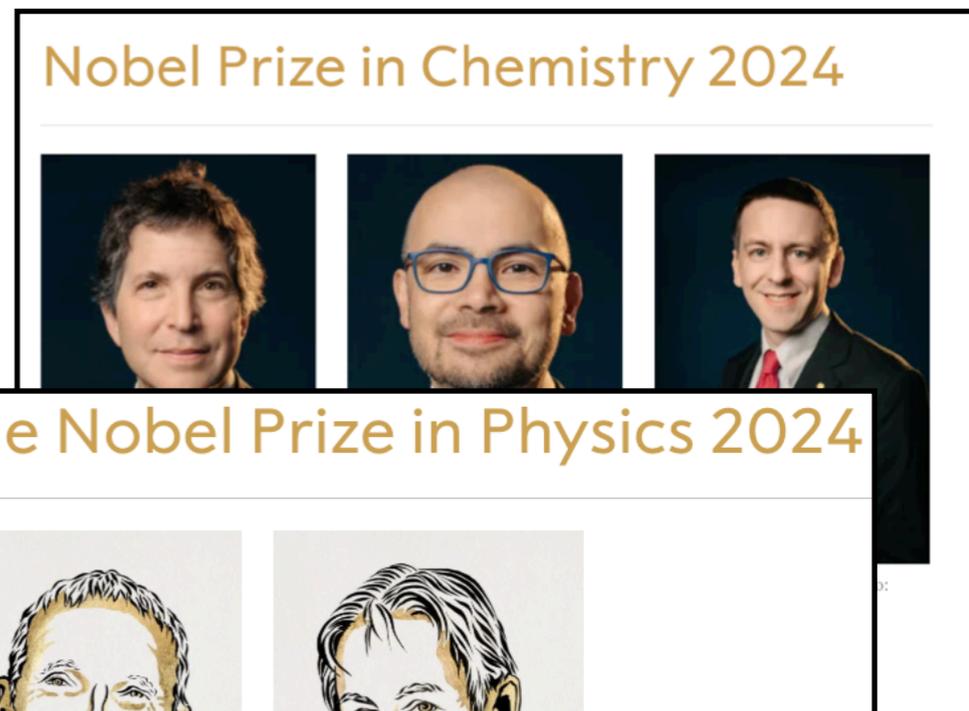
CERN did play a role in large-scale
development **beyond HEP**

State of AI in Physics



200+ papers in 2024

Similar for nucl-ex,
astro-ph, hep-ph, ...



Rapid **rise of AI** in fundamental physics

Transition from **concepts to applications**

Recognition of AI work as **Nobel-prize worthy**

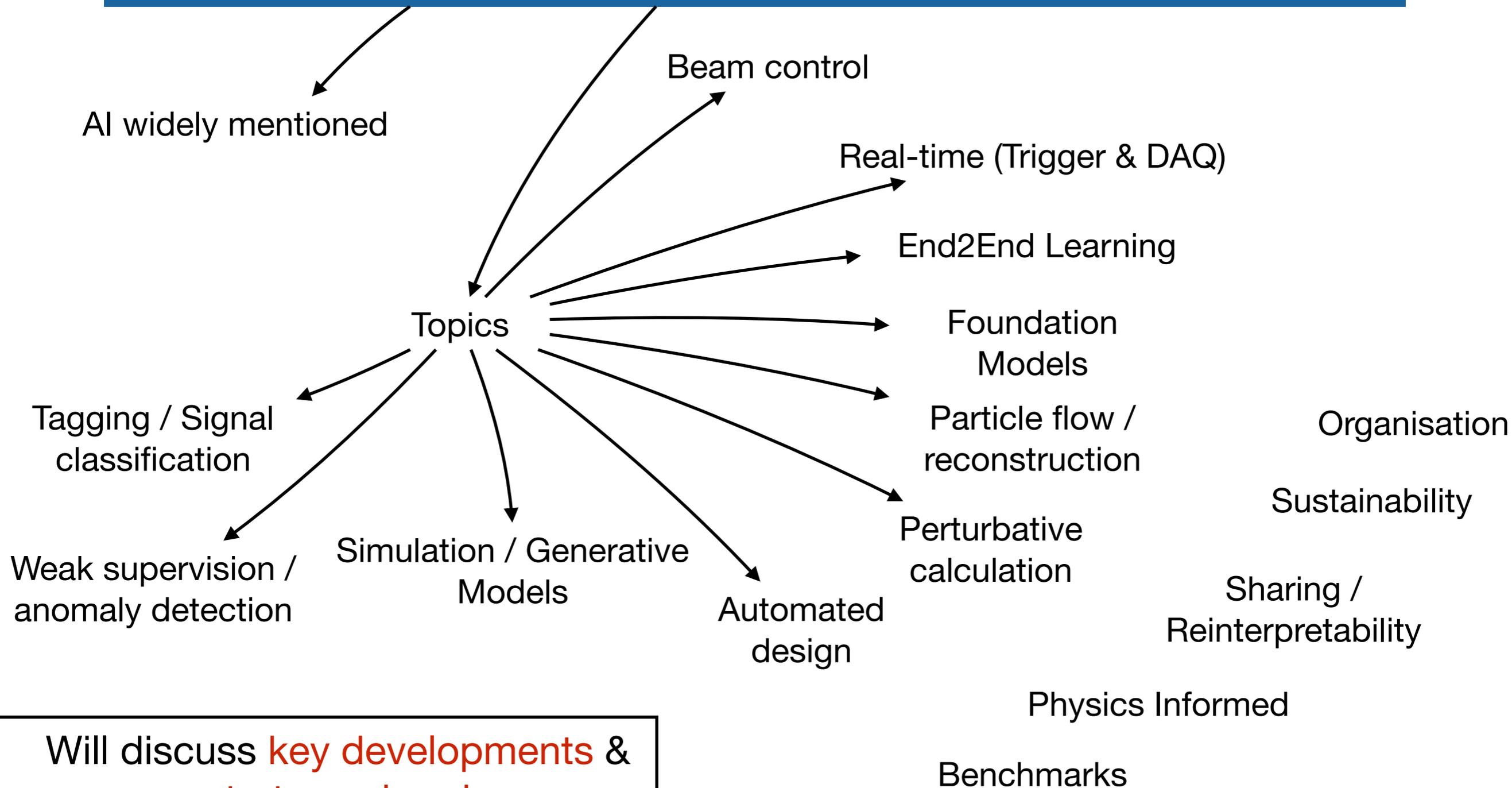


Input to the European Strategy for Particle Physics - 2026 update

AI widely mentioned



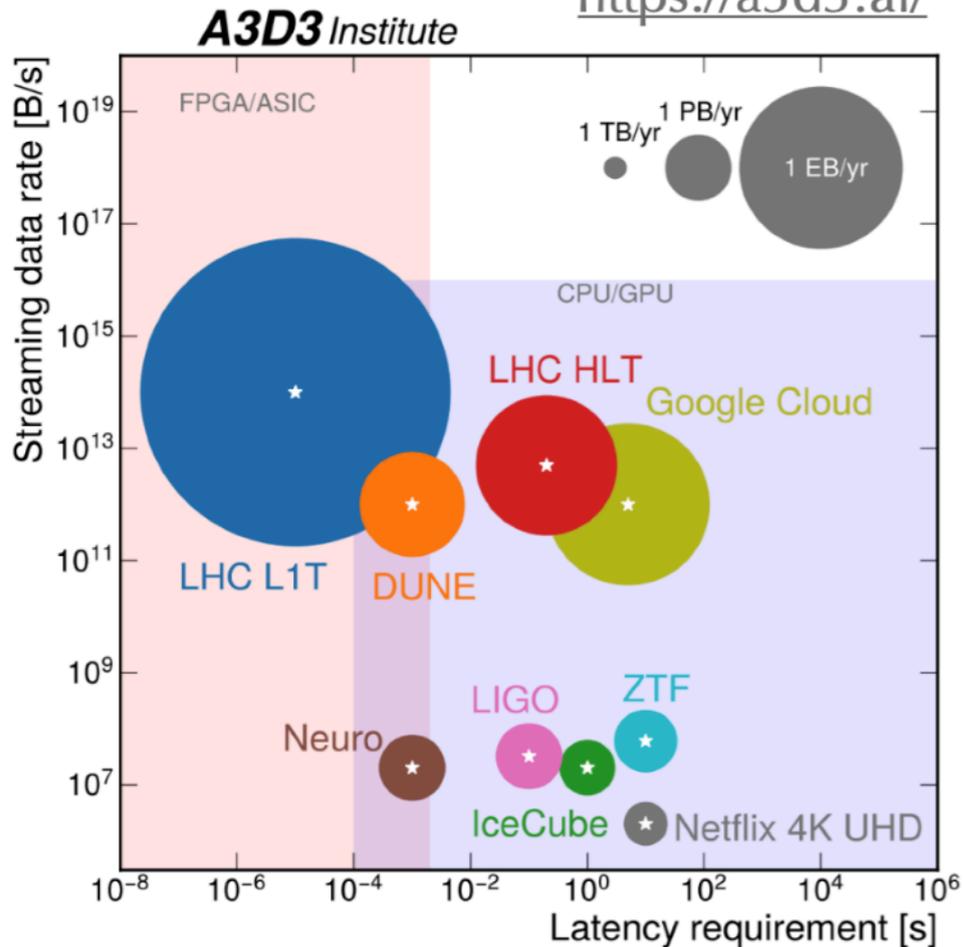
Input to the European Strategy for Particle Physics - 2026 update



Will discuss **key developments** & **strategy ahead**

Data Taking

<https://a3d3.ai/>



Collider physics offers combination of **data rate and latency** requirements

Relevant for data reduction and read-out

Need efficient **ML on FPGA** hardware for **current and future experiments**

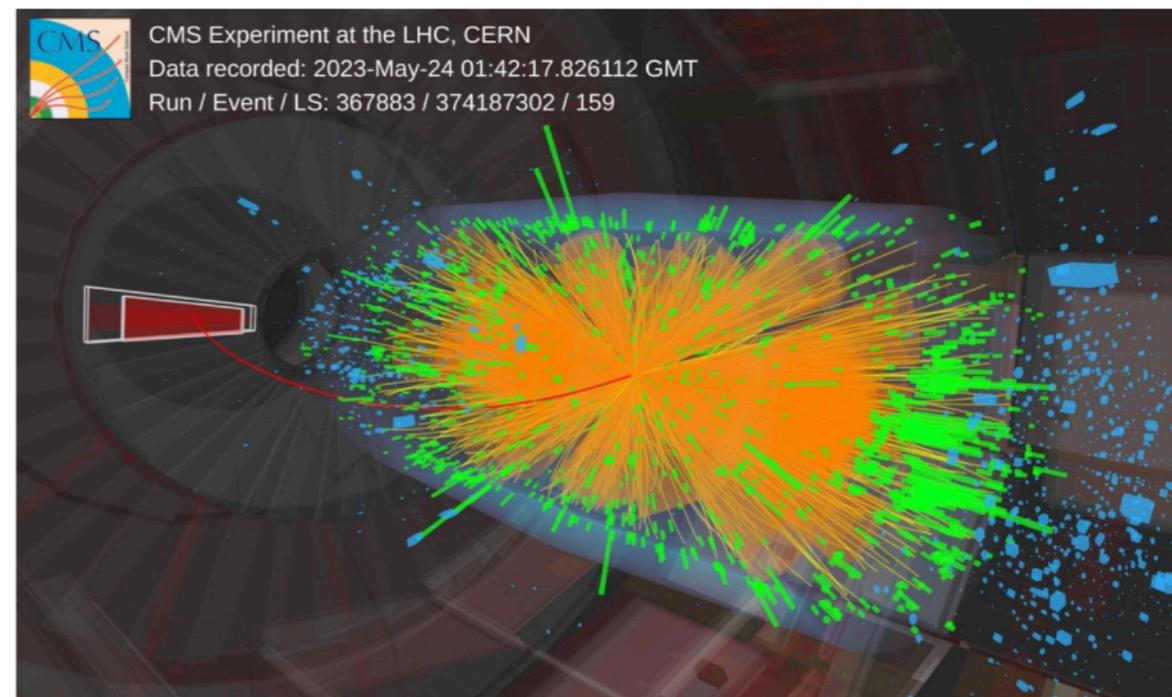
More in Thea's talk

Already now: Exploring **anomaly** detection via autoencoders in **CMS L1**



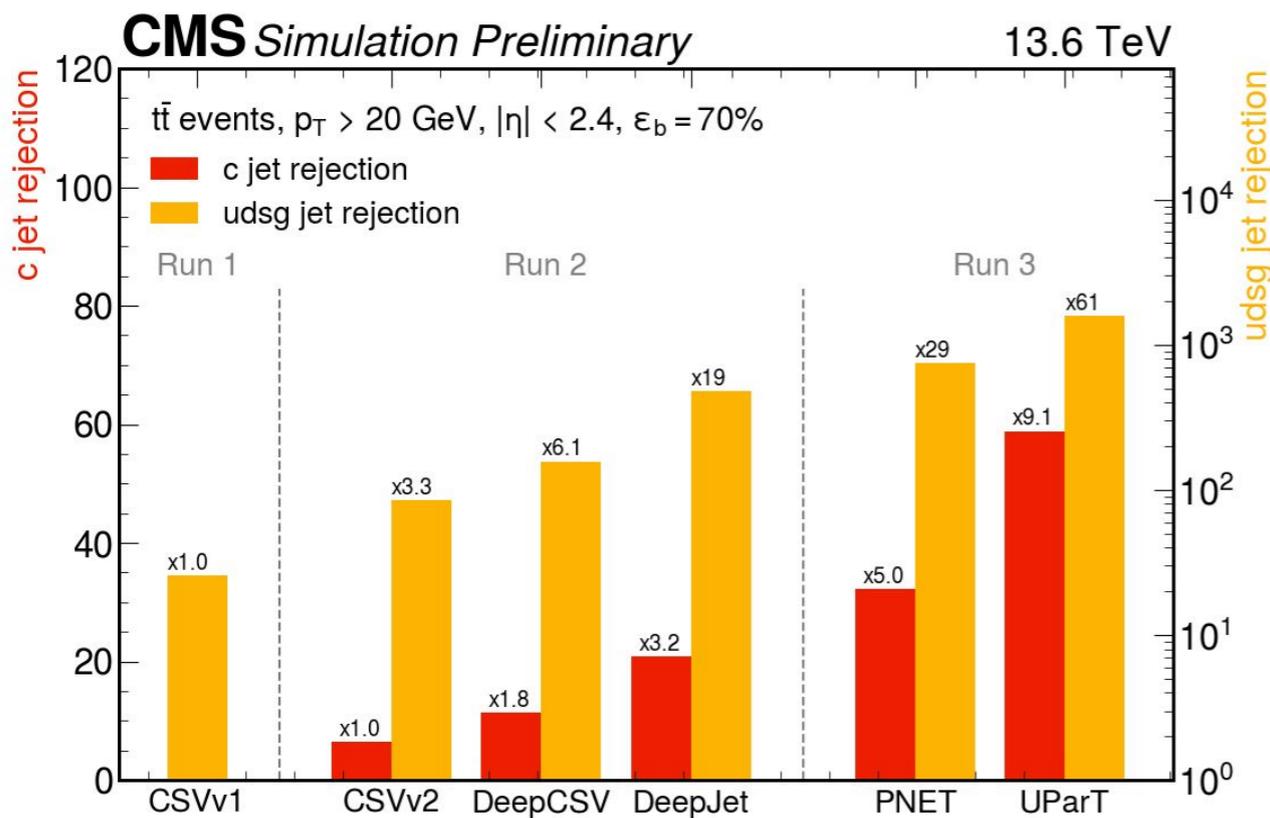
AXOLITL

hls 4 ml



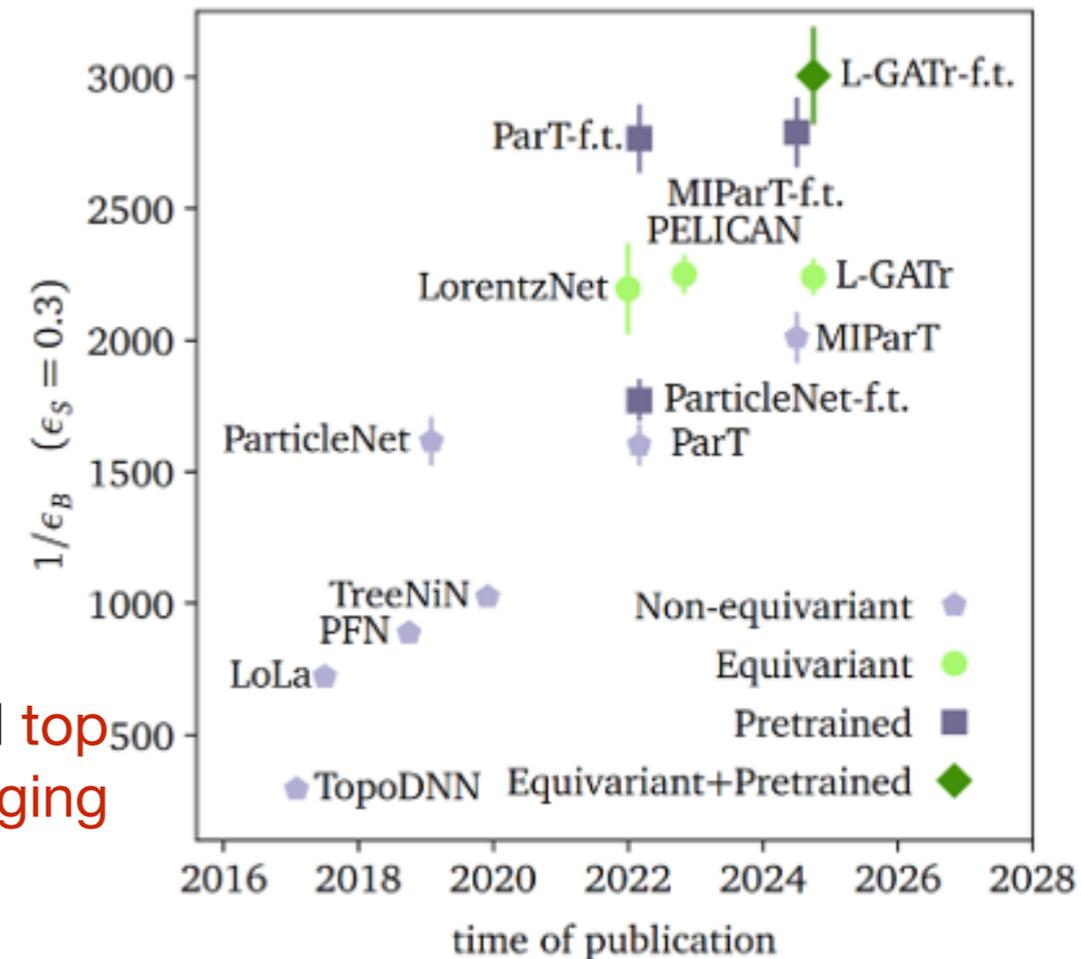
Tagging

- Determine source of a jet/event/..
- **Wide-spread** application of **classification**
- **Flavour/charm** tagging relevant **Higgs factories**



Flavour tagging

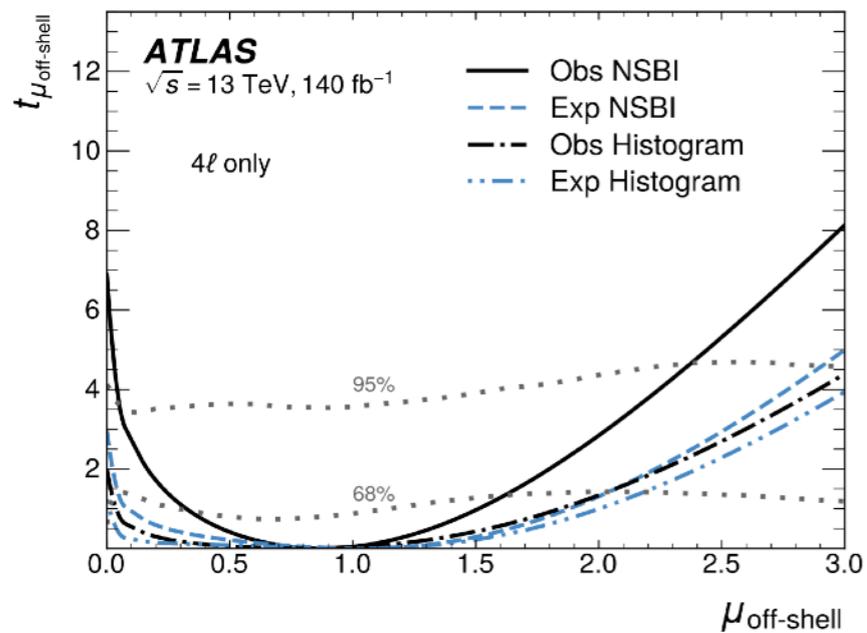
Boosted top quark tagging



Orders of magnitude improvement in key physics application from better AI

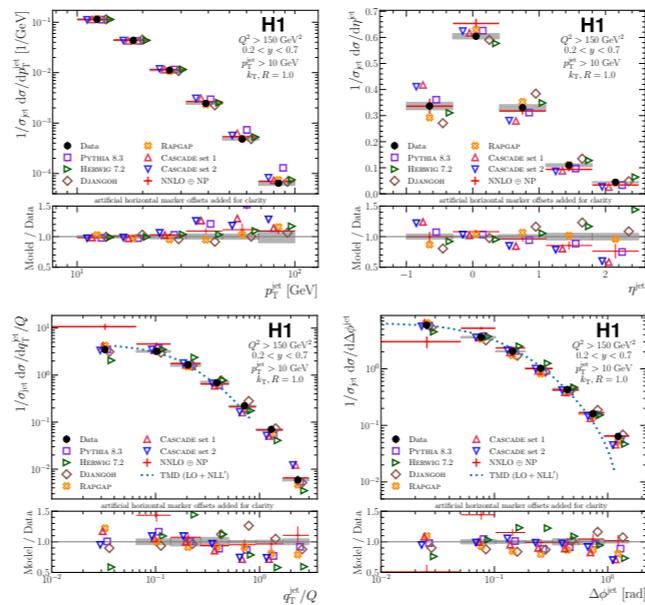
Direct pipeline from AI development to deployment

Analysis Paradigms



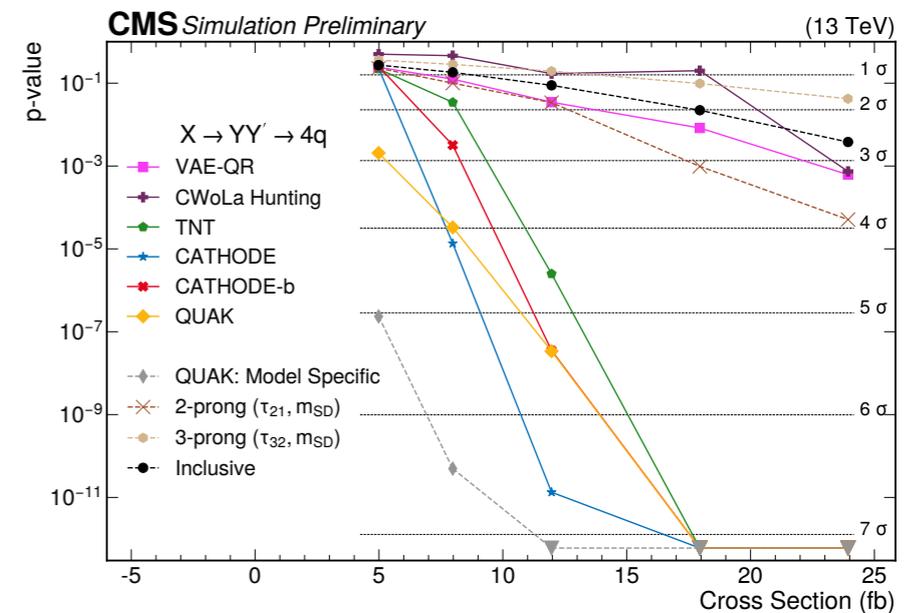
Simulation Based Inference

Unbinned high-dim shape information in statistical analysis improves sensitivity



Unfolding

Unfold in higher dimensions via reweighting and morphing



Anomaly detection

New searches with reduced model dependency

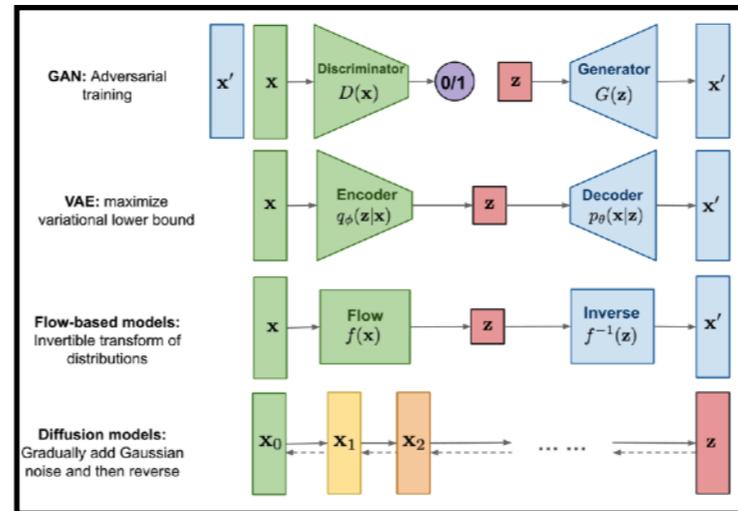
Substantial improvements and qualitatively new approaches done already **now**

Simulation

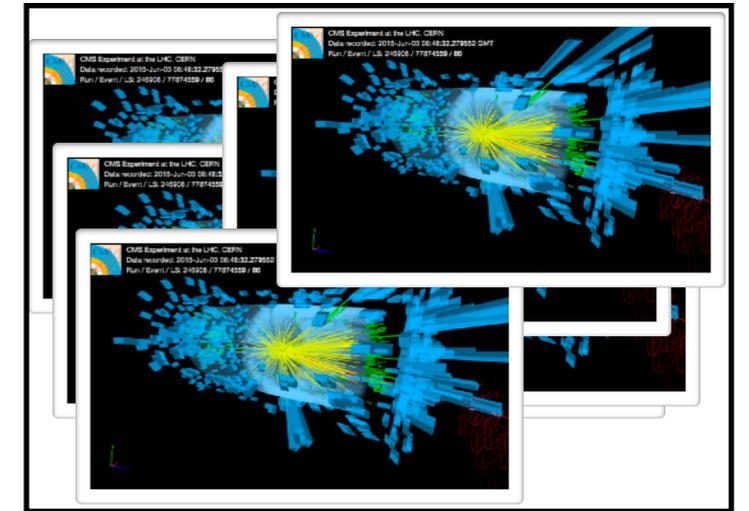
1. Simulation or collider data as input



2. Train generative surrogate



3. Sample



Fast Calorimeter Simulation Challenge 2022

Hardware	Simulator	NFE	Batch Size	Time / Shower [ms]	Speed-up
CPU	GEANT4			3914.80 ± 74.09	$\times 1$
	CALOCLOUDS	100	1	3146.71 ± 31.66	$\times 1.2$
	CALOCLOUDS II	25	1	651.68 ± 4.21	$\times 6.0$
	CALOCLOUDS II (CM)	1	1	84.35 ± 0.22	$\times 46$
GPU	CALOCLOUDS	100	64	24.91 ± 0.72	$\times 157$
	CALOCLOUDS II	25	64	6.12 ± 0.13	$\times 640$
	CALOCLOUDS II (CM)	1	64	2.09 ± 0.13	$\times 1873$

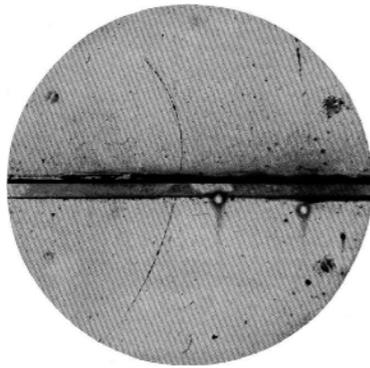
Broad effort to **speed-up / bypass simulation at detector** (e.g. calorimeters) **and theory** (event generators MadNIS/ Sherpa/..) level

~2 orders of magnitude speed-up on same hardware achieved

Essential for simulation needs of next generation experiments

1712.10321, 2109.02551, 2309.05704, 2410.21611, ...

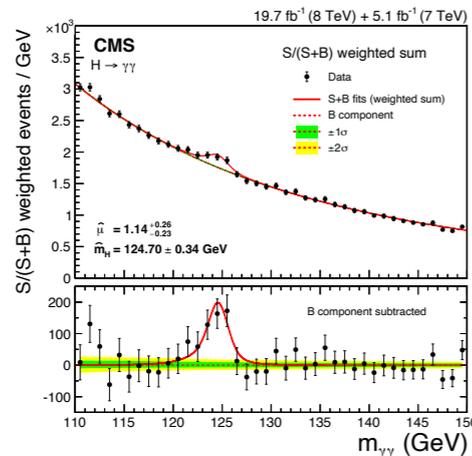
End to end



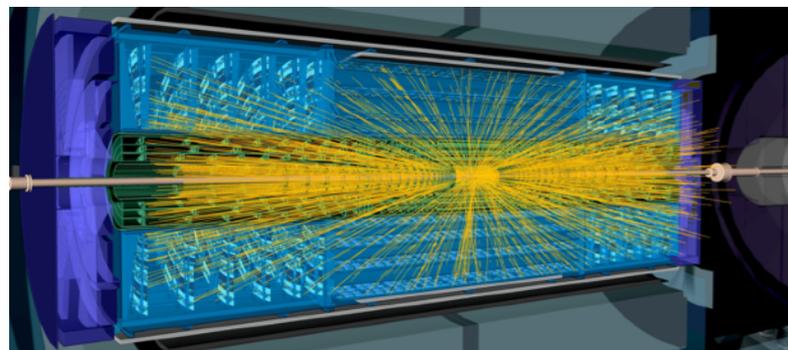
1930s — Positron discovery: Single event

Channel:	SVX
observed	27 tags
expected background	6.7 ± 2.1
background probability	2×10^{-5}

1990s — Top discovery: Multiple events, cut & count



2010s — Higgs discovery: Histogram & shape information



20??s — X discovery: Holistic processing of low-level information

Separately optimising

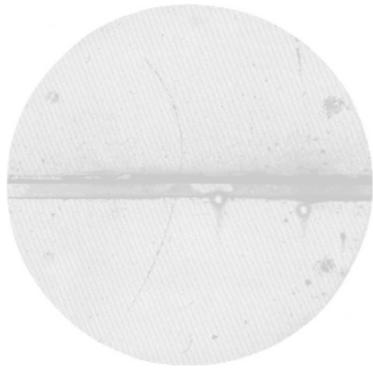
...detector calibrations, particle flow objects, jet reconstruction, pile-up suppression, object tagging, background rejection, signal selection...

loses information at each step

Future analysis will be event-level and end-to-end for optimal sensitivity

Jets and other objects for human consumption only

End to end



1930s — Positron
discovery: Single event

Separately optimising

...detector calibrations, particle flow
objects, jet reconstruction, pile-up
suppression, object tagging, background
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loses information at each step

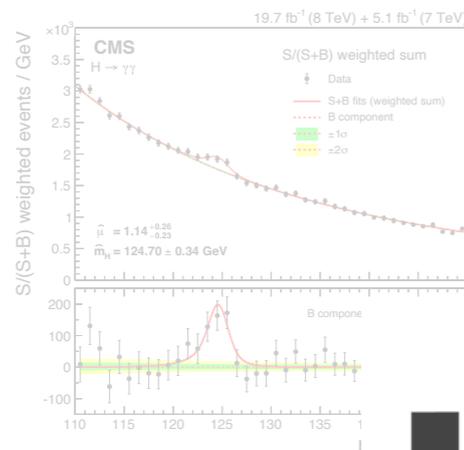
Future analysis will be event-wise
and end-to-end for optimal
sensitivity

Jets and other objects for human
consumption only

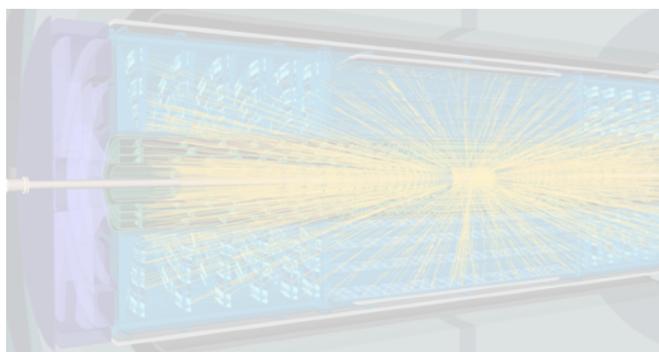
One step further: Differentiable
programming and surrogates
allow end-to-end optimisation of
detector design

Channel:	SVX
observed	27 tags
expected background	6.7 ± 2.1
background probability	2×10^{-5}

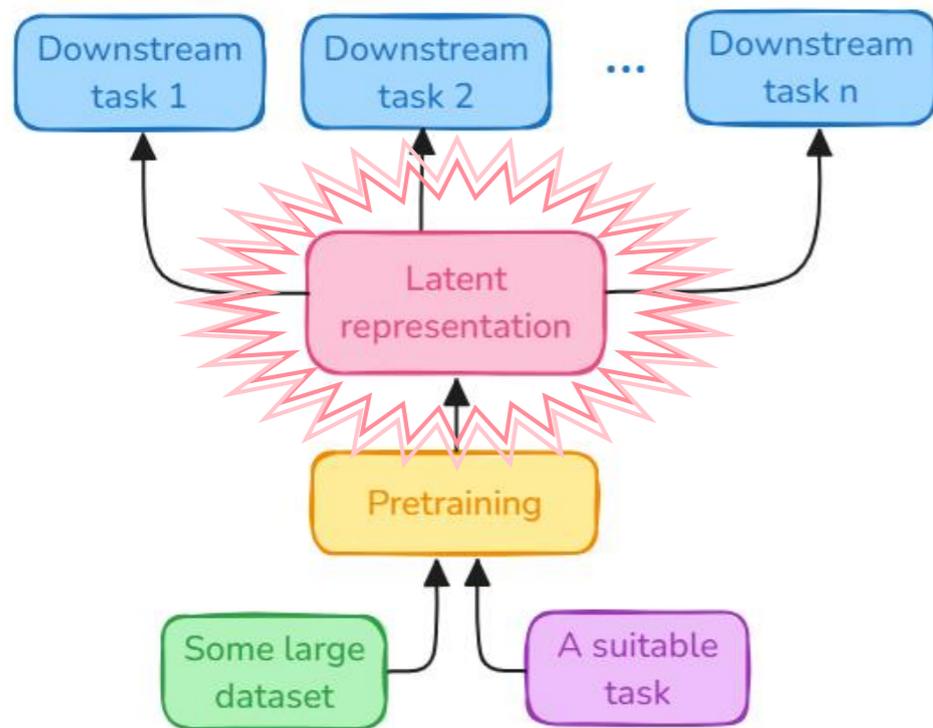
1990s — Top discovery:
Multiple events, cut &
count



2010s — Higgs discovery:
Histogram & shape
information



Foundation models

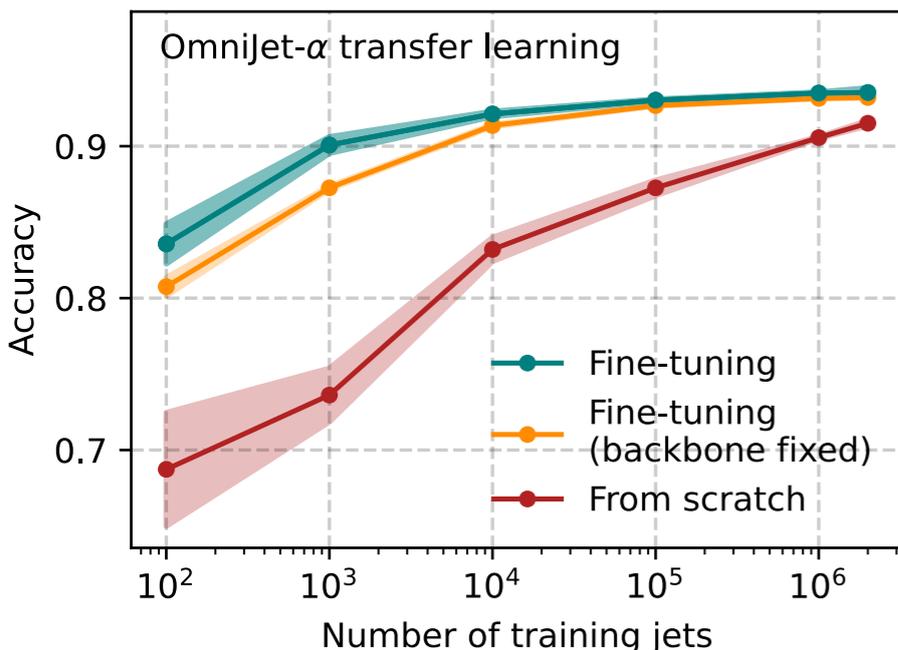


Similarity across many AI problems in fundamental physics

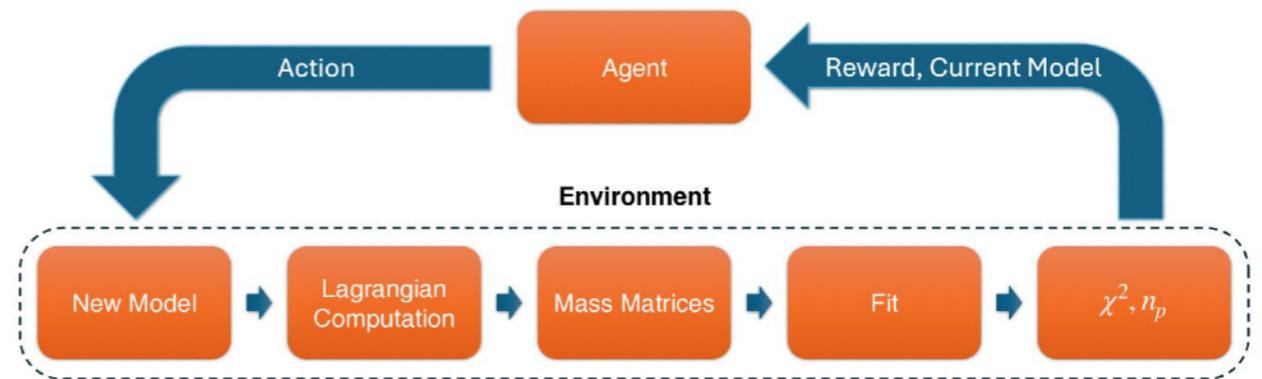
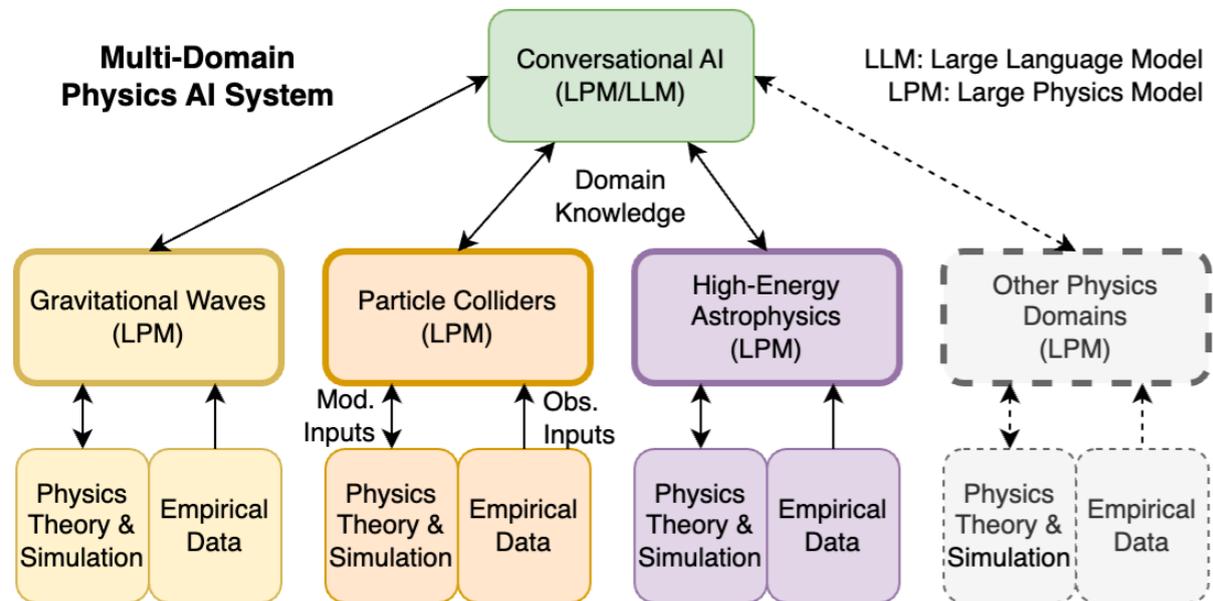
Build multi-dataset multi-task foundation models that can be adapted via **fine-tuning to new problems**

Learn underlying **joint representation**

Improve:
performance (**higher max. accuracy**)
data efficiency (**lower number events needed**)
scientific turn-around (**lower time to result**)



Large-physics models



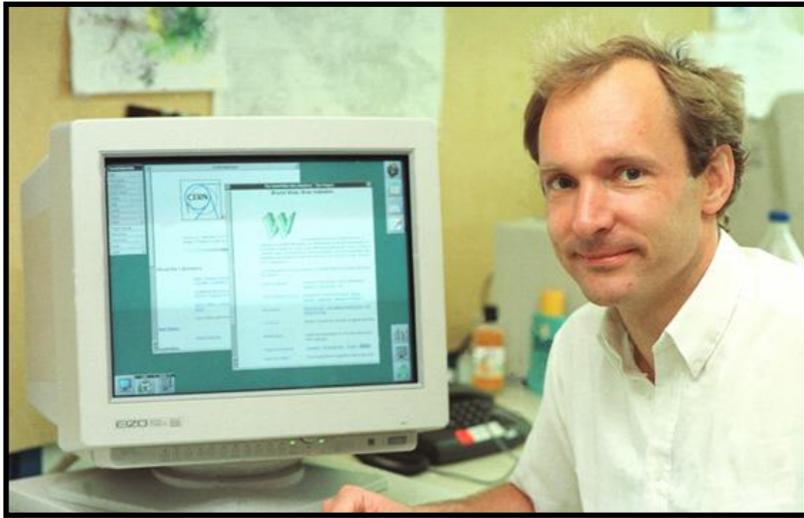
Language models as glue
between domain experts

Agent-based hypothesis
generation & iteration

Advances in large language models (LLMs) allow a
new degree of autonomous operation

Leverage for information systems, accelerator control,
automated analysis design, ...

Physics for AI



- Problems from **fundamental physics** historically **driving** force for **informatics**
- Chance to **regain** with **future collider** strategic push
- Topics of:
 - real-time challenge
 - multi-source uncertainty quantification
 - ultra-rare signals
 - complex detector data
 - ...

The Nobel Prize in Physics 2024

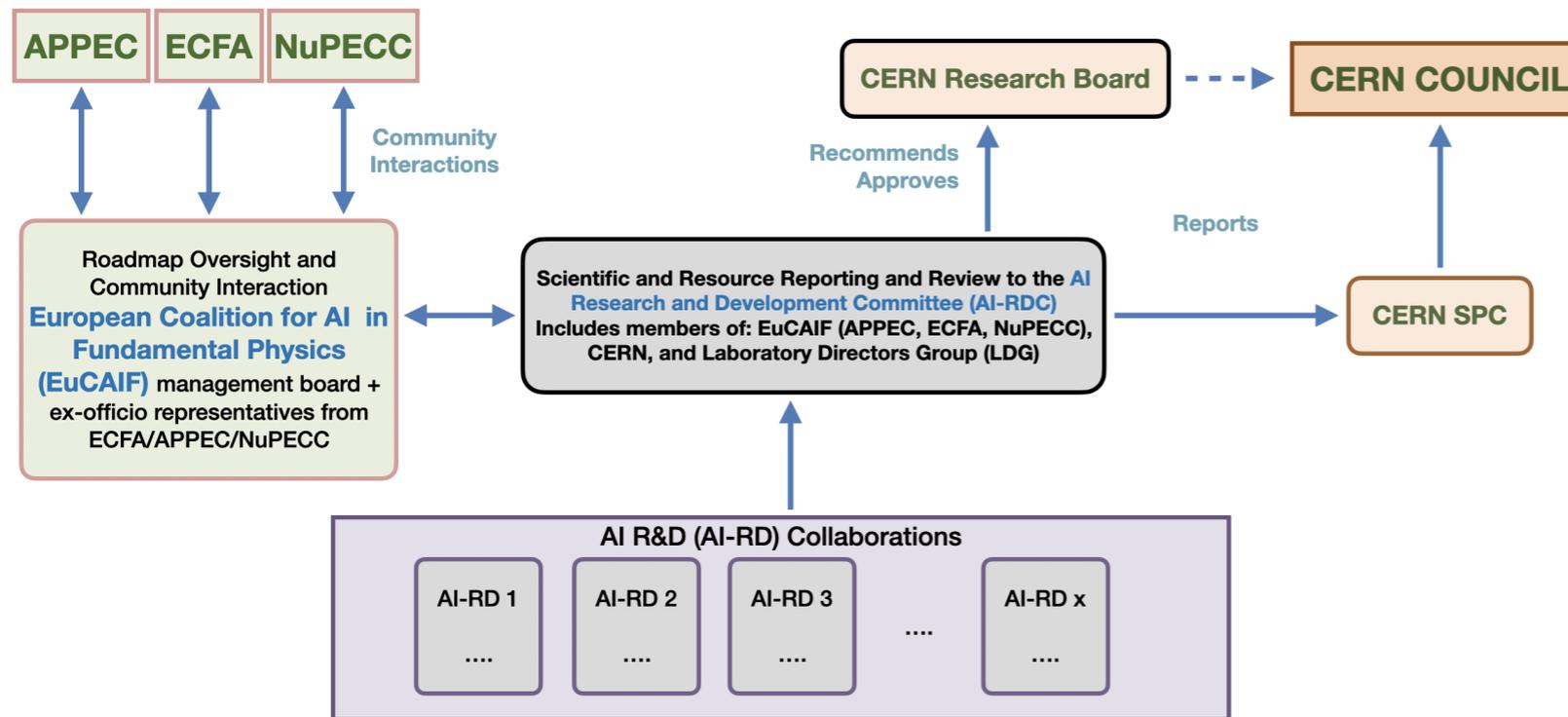


III. Niklas Elmehed © Nobel Prize
Outreach
John J. Hopfield
Prize share: 1/2



III. Niklas Elmehed © Nobel Prize
Outreach
Geoffrey Hinton
Prize share: 1/2

How to get there?



- Form and coordinate **AI R&D collaborations**
 - Organise **research**
 - Central provider of **tools**
 - Training and **career development**
- AI-RDs as **nucleus for a future flagship AI** effort

Examples:

- AI for Data Processing
- AI for Detector and Accelerator Control
- AI for Detector Optimization
- AI for Event Reconstruction
- AI for Analysis
- AI for Generation and Simulation
- AI for Theory
- AI for HPC Usage
- AI for Documentation and Education

Environment



AI and basic research in the natural sciences

Artificial intelligence (AI) is of central importance for basic research in the context of the framework programme “Exploration of the Universe and Matter – ErUM”.¹ It opens up entirely new dimensions of research in terms of speed, precision, and scalability, thereby enabling scientific questions to arise that were previously unthinkable. At the same time, it is indispensable for handling unprecedented amounts of data, such as those generated by large-scale research infrastructures. The data-intensive nature of basic research in the natural sciences plays a key role in driving the development of new AI algorithms and would not be possible without AI methods. Scientific requirements for data analysis and statistics are fostering the creation of innovative methods, particularly in the field of explainable AI, where results must be traceable and interpretable.

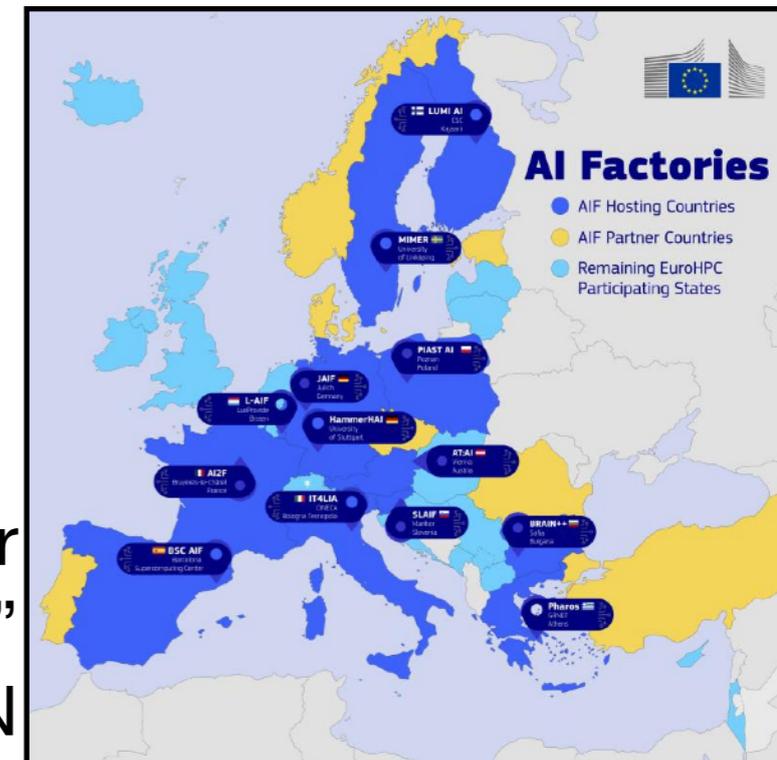
At the same time, AI systems greatly benefit from the high-quality, non-personal datasets provided by ErUM disciplines—such as those from particle collisions or sky surveys—which offer ideal training resources. These models can detect patterns even in low-resolution data, a capability that is also highly relevant for fields like cancer diagnostics. The close interplay between fundamental research and AI development not only advances science but also yields domain-specific AI solutions that are particularly relevant for industrial applications.

To fully leverage the potential of AI for basic research in the natural sciences in Europe, the following priorities are essential:

- AI funding should specifically promote interdisciplinary approaches and foster the exchange of solutions across disciplines to accelerate cross-learning and unlock synergies (for example, funding initiatives from Big Data to Smart Data in ErUM-Data or Mathematics for Innovations). Particularly promising are versatile AI approaches with the potential to accelerate multiple research domains simultaneously.
- Furthermore, initiatives focused on networking (e.g. EuCAIF at European level, or ErUM-Data-Hub at national level) and on developing explainable AI should be strengthened. Establishing structured and reliable pathways for the transfer of research results into commercial applications is also crucial and should be strategically addressed.

German input to European AI strategy highlights particles collisions

Build conditions for future “AI Factory” at CERN



Link to future European large-scale AI initiatives

Politics expects leadership on AI in science from us

→ Benefit from CERN for AI discussion

Opportunity from possible future collider project as first AI first machine

Conclusions

- Substantial responses on AI
- Breakthroughs in AI directly translate to **better physics**
 - Qualitative: Powerful **new analysis paradigms**
 - Quantitative: **Multiple order-of-magnitude** improvements
- Chance to build future generation experiments **AI first**
 - **End-to-end** designed for sensitivity
 - Consider in planning **physics reach**
 - Be open to **continued innovation**
- **Boosts** overall **future collider effort**
 - Our **community** is expected to **lead in AI**
 - Attract **innovations** and **AI funding**
- Start now with substantial effort
 - Form **AI R&D collaborations**
 - Use HL-LHC as testbed

Thank you!