## New proposal: R&D for the exa-scale computing environment

Team:

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### Introduction

- This new proposal follows COMP\_04 (and COMP-03) as its successor
- **COMP\_04**:
  - "Evolution of the computing environment for high-energy and astro-particle experiments"
  - The project was led by Fabio Hernandez, Renaud Vernet, Sebastien Gadrat, and Tomoaki Nakamura from FY2019 to FY2024
  - The close collaboration between IN2P3 Computing Center (CC-IN2P3) and KEK Computing Research Center (KEK CRC) in computing

 $\rightarrow$  Let me report the achievements of COMP\_04 in FY2024





### CC-IN2P3 and KEK CRC

- CC-IN2P3 and KEK CRC support a wide range of experimental projects
  - Including Bellell, LHC, T2K, etc
- > Many common challenges to overcome
  - Both hardware and software areas
- > Shared interests regarding new technologies
  - > Machine learning(ML), ID federation, etc



### → Exchange of experience, and know-how is quite essential !!



### Yearly F2F workshop

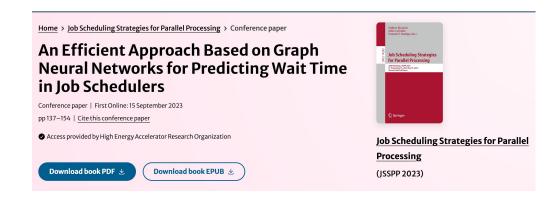
- F2F workshops are unique opportunity to discuss on-going work and new technology at both centers
- > Topics of 2025 workshop@Lyon were:
  - Status report from CC-IN2P3 and KEK CRC
  - News from LSST experiment
  - Open access project at KEK

### https://indico.in2p3.fr/event/35206/



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### Al session

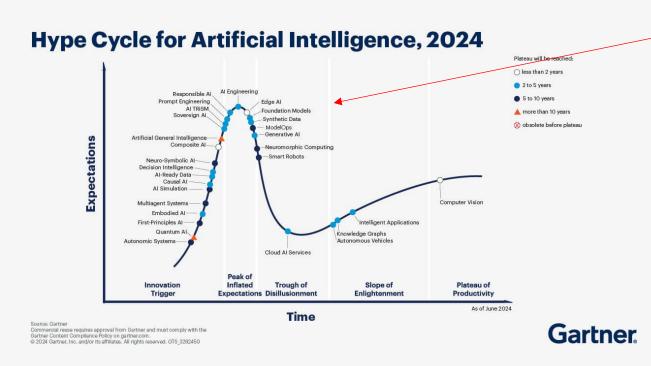


WED, FEBRUARY 19				
<b>8:30 AM</b> → 9:00 AM		Welcome coffee	<b>③</b> 30m	
9:00 AM → 10:00 AM	Al session 9:00 AM	Large Langage Model Speaker: Mr Mehdi Hennequin	©1h	
<b>10:00 AM</b> → 10:30 AM		Coffee break	<b>③</b> 30m	
<b>10:30 AM</b> → 12:00 PM	Al session	A study of foundation models for event classification in collider physics Speaker: Tomoe Kishimoto (High Energy Accelerator Research Organization)	𝔅 40m	
	11:10 AM	MLLMs in Scientific Research and High-Performance Computing Speaker: Imed MAGROUNE	𝔇 50m	

- ML has been a focus of joint collaborative efforts between CC-IN2P3 and KEK CRC
  - A successful application for estimating job waiting time in batch schedulers
  - → Published in Lecture Notes in Computer Science
- New application has been investigated:
  - The idea is to apply the concept of foundation models to collider physics
  - A potential approach to reduce computing resource usage in future collider experiments



### Foundation models



- "Foundation models" is one of the keywords for AI
  - Pre-training using a larger amount of "unlabeled" data
  - Fine-tuning for a target application (transfer learning)

→ Q: Is the concept of foundation models beneficial to collider physics?

Gartner.com

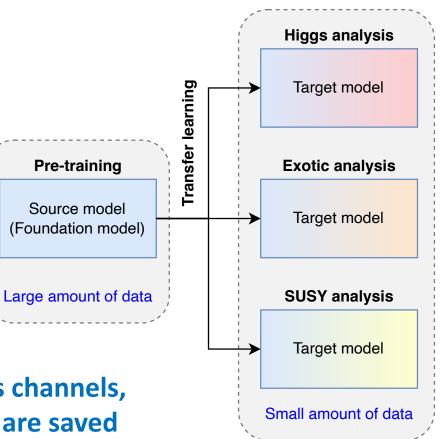
加速器だから見える世界。 **KEK** 

### Use case of physics analysis

> Many analysis channels in collider physics

- Higgs, Exotic, SUSY, etc
- Dedicated Deep learning models are trained from scratch for each analysis channel
- Large amount of training data (MC) for each channel

→ If transfer learning can be applied to different analysis channels, computing resources for MC simulations and DL training are saved



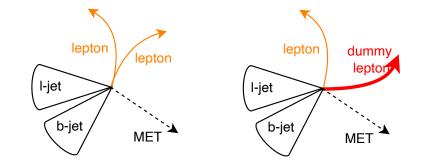
### Pre-training strategy

> Only low-level features of each object (4-vector+chargte) are used as inputs

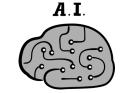
Real collision data (CMS open data) are used in pre-training

> Self-supervised learning is employed to handle the unlabeled data

- Pre-training strategy
  - An object is randomly replaced with a dummy object when preparing mini-batch



 $\rightarrow$  DL model is trained to predict what type of object was replaced

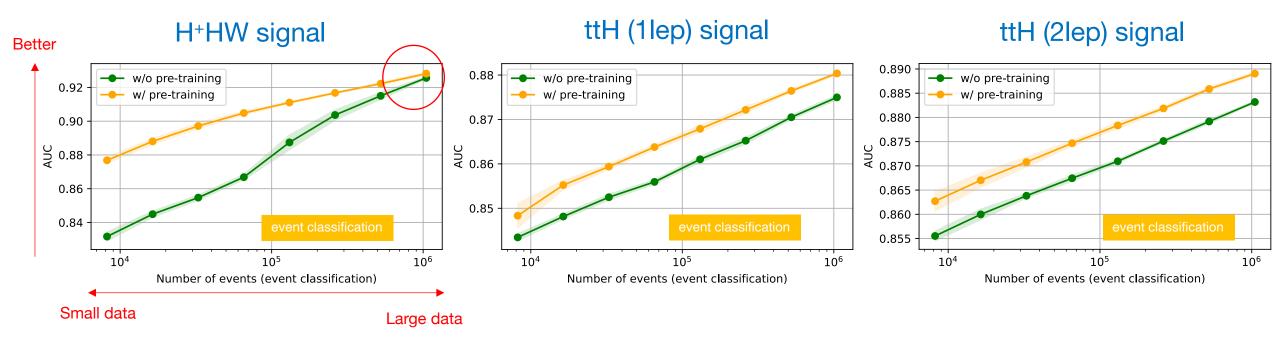


...lepton is dummy(?)



### AUC of event classification

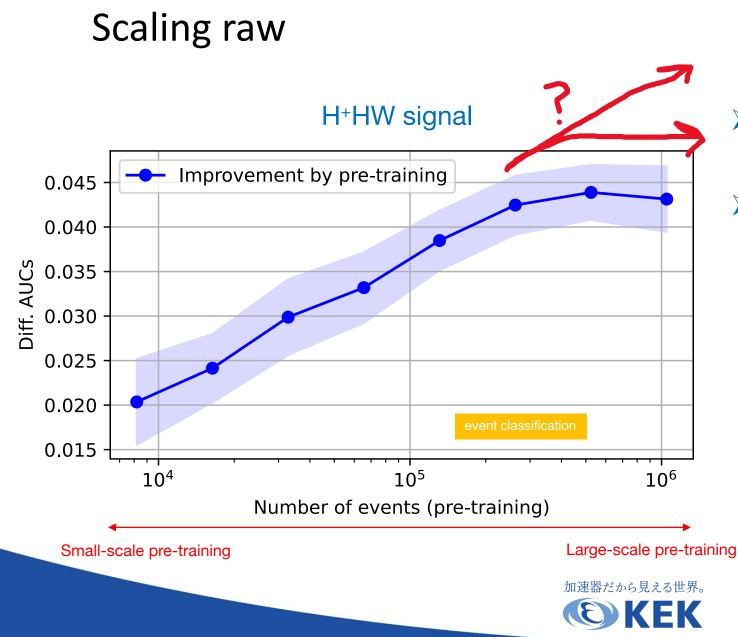
> The concept is examined using "Event classification" problem : signal vs background



The improvements are confirmed for different signal events

→ The pre-trained model (foundation model) generalizes well

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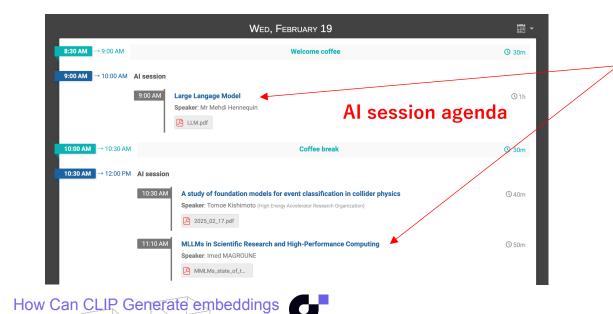


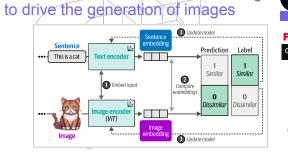
The scaling behavior with respect to the number of events in pre-training

### Future works:

- Need to check with larger models and larger data
- Investigate new data augmentation approach
- Explainability of AI: what kind of knowledge is learned during pretraining and utilized in downstream tasks...?

### Large Language Models





From Foundation models to smallers LLMs

### Jetson Orin Nano Developer Kit NVIDIA DeepSeek-R1-Dis Dolt.city Where ideas begin



Dres3.1-405B-Instruct and GPT-40 on MATH-500 and AIME24, (b) & (c) DeepSeek-R1-Distill-15: outperforms of preview on MATH-500 and AIME24, and supposes ol-min on MATH-500; (c) & (f) DeepSeek-R1-Distill-7B beats ol on MATH-500 and AIME24, and exceeds DeepSeek-R1 on AIME24

Work done during an internship at Shanghai Al Laboratory Corresponding authors: Biqing Qi (qibiqing@pjlab.org.cn), Bowen Zhou (zhoubowen⊜tsinghua.edu.cn)

- We also discussed Large Language Models (LLM) and their applications in our fields during the workshop@Lyon
- User support for the efficient use of computing resources is a key challenge shared by CC-IN2P3 and KEK CRC
  - Distributed computing, HPC cluster, containerized workflow, Networks, etc

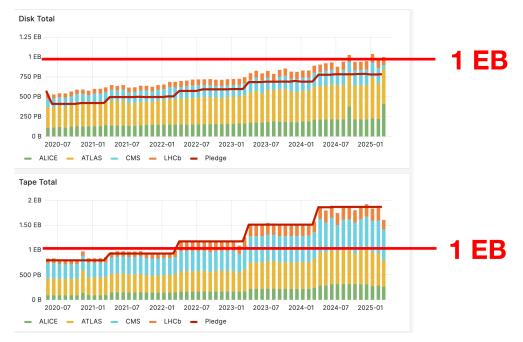
→ We are planning to start a research project on chatbots for user support



### New proposal

- Multiple approaches for introducing brand-new technologies are necessary for the evolution toward the exa-scale computing environment
- Refine and extend the knowledge gained from COMP\_04
- > 3 topics selected for this year:

### **LHC** experiments



- Study of foundation models for event classification in collider physics
   → This topic has been discussed in previous slides
- 2. Research and application of chatbots (based on LLMs) in High Energy Physics
- **3.** Research on enhancing the convenience of IT services based on identity federation



### Chatbots in HEP

- User support in data center operations requires access to internal information
  - > System configuration, Past tickets, etc
- Normal chatbots can't answer such questions (of course!)

# → Retrieval-Augmented Generation (RAG) is a promising approach

ChatGPT 4o >

Could you please provide the IP address range for KEK's DC cluster?

The specific IP address range for KEK's Data Center (KEKCC) cluster is not publicly disclosed in the available documentation. However, access to KEKCC resources is managed through designated servers: web2.INFN.IT

- Access Servers: sshcc1.kek.jp and sshcc2.kek.jp
- Work Server: login.cc.kek.jp web2.infn.it KEKCC.KEK.JP

These servers facilitate secure connections for authorized users. Notably, direct access to the work server (login.cc.kek.jp) is restricted to clients with registered IP addresses. This implies that KEK maintains a list of approved IP addresses for security purposes. To gain access, users typically need to submit their machine's IP address and hostname for registration. This process is outlined in KEK's support documentation. INDICO.BELLE2.ORG REKCC.KEK.JP WEB2.INFN.IT

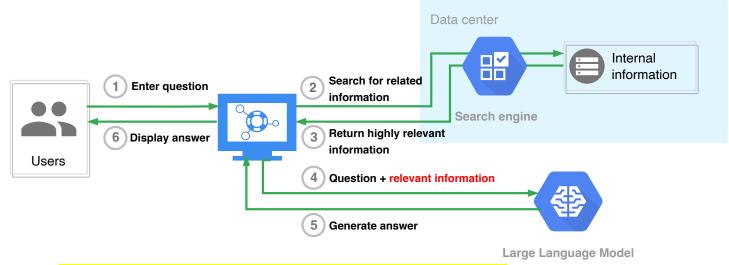
If you require the exact IP address range for network configuration or firewall settings, it is recommended to contact the KEK Computing Center directly or consult with your KEK project liaison. They can provide the necessary information tailored to your access needs.





### RAG

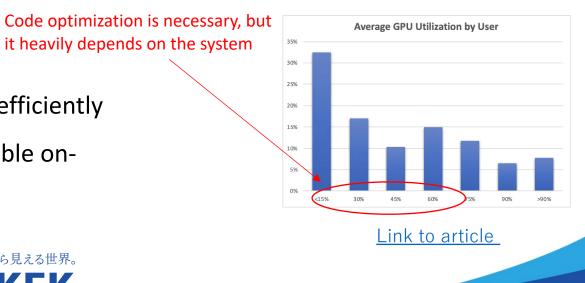
- Relevant information is retrieved and fed into LLMs for processing
  - Official documents, past ticket, e-mails, etc



→ Develop a chatbot for user support, including code generation optimized for our computing infrastructure

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- Related research topics:
  - > Methods for searching relevant information efficiently
  - Methods for fine-tuning or distillation to enable onpremise deployment of LLMs



### **ID** federation

- ID federation plays a crucial role in the efficient deployment of IT services (including chatbot)
  - Users are able to use services provided by other institutions with the credentials from their home institution
    - → Reduce the effort required for account management
- > Academic ID federation:
  - $\geq$  European ID federation  $\rightarrow$  eduGAIN
  - $\succ$  Japanese ID federation  $\rightarrow$  GakuNin
  - GakuNin can federate with eduGAIN

→ Plan to deploy a test service for verifying authentication and the exchange of user attributes

ReduGAIN 🚳 GakuNin



### Migration: X.509 to Token

ID federation is particularly critical for "Grid computing"

- The Grid authentication infrastructure is migrating from X.509 certificates to tokenbased authentication
  - > KEK CRC is deploying a token issuer service (INDIGO IAM) for Bellell experiment
  - > But, a user registration procedure for INDIGO IAM is no established yet...

→ Managing user identities and accounts for all collaborators with the security level required by Grid computing <sup>(\*)</sup> is a significant challenge

→ We need ID federation!

(\*) Identity Assurance Level2 must be referred to F2F meeting or human resource DB



### Summary

> Information exchange is highly valuable for both centers

- In the new proposal, we plan to focus on the application of AI technologies and ID federation
  - Milestones for FY2025: Development of an initial chatbot prototype and deployment of a test service for ID federation
- > To strengthen the partnership, we request funding for:

A new initiative

- ➤ KEK-CRC → IN2P3-CC visit in Spring to exchange information on computing technologies (annual workshop)



## Backup



## Datasets (CMS Opendata)

		Selections	# of events	
Pre-training —	Collision data	lepton $\ge 1 + jets \ge 2 + bjets \ge 1$	~10 <sup>6</sup>	
Event	H+tb[ <u>ref.]</u> vs ttbar+jets	lepton $\geq$ 1 + jets $\geq$ 4 + bjets $\geq$ 1	~10 <sup>6</sup>	
	H+HW[ <u>ref.</u> ] vs ttbar+jets	lepton $\ge 1 + tau \ge 1 + jets \ge 3 + bjets \ge 1$	~10 <sup>6</sup>	
	ttH[ <u>ref.</u> ] vs ttbar+jets	lepton $\ge 1 + jets \ge 4 + bjets \ge 2$	~10 <sup>6</sup>	
	ttH[ <u>ref.</u> ] vs ttbar+jets	lepton $\ge 2 + jets \ge 2 + bjets \ge 1$	~10 <sup>6</sup>	

Pre-training is performed using collision data (unlabelled data) based on the foundation model concept

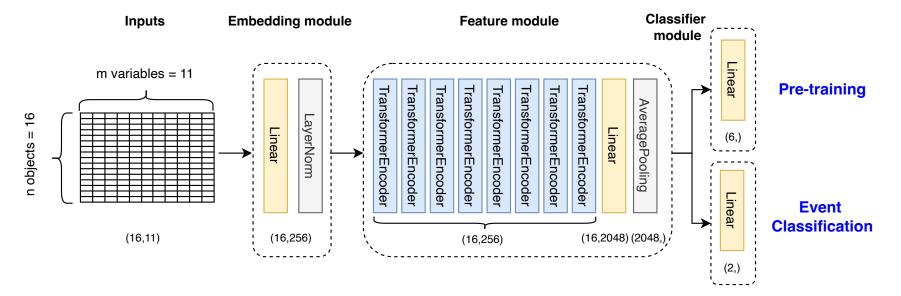
 $> \sim 10^7$  events are available after the selection, but only  $\sim 10^6$  events are used

> NVIDIA A100: ~10<sup>4</sup> events/sec (10<sup>7</sup> events /10<sup>4</sup> x 500 epochs = 138 hours)



## DL model

- > Transformer encoder is employed:
  - ~11M trainable parameters



 $\rightarrow$  Weight parameters of embedding and feature modules are transferred and fine-tuned  $\rightarrow$  Classifier module is always trained from scratch

