

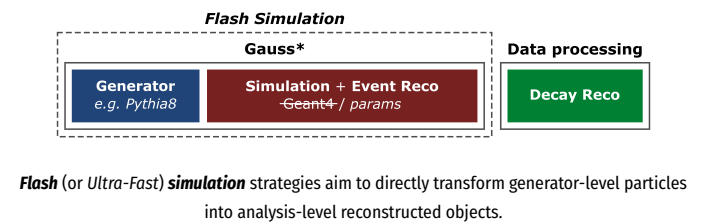
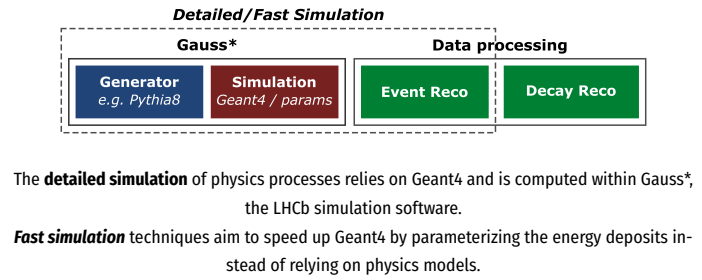


## 1. Motivation

**Detailed simulation** of the interactions between particles and the LHCb detector requires significant CPU resources.

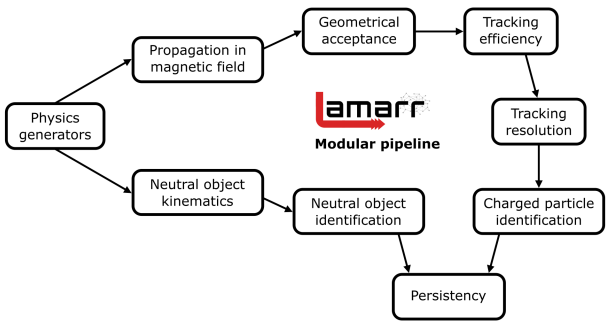
- LHCb has used **over 90% of CPU resources** for simulation during Run 2;
- Current approaches do **not scale** to future needs;
- Must develop **faster simulation options!**

## 2. Fast simulation VS. flash simulation



## 3. What is Lamarr?

**Lamarr** is the novel flash-simulation framework of LHCb, able to offer the fastest option to produce simulated samples. Lamarr consists of a **pipeline of** (ML-based) **modular parameterizations** designed to replace both the simulation and reconstruction steps.



The Lamarr pipeline can be split in two branches:

- charged particles** require tracking and particle identification models;
- neutral objects** need to face the *particle-to-particle correlation* problem.

## 4. Models under the $k$ -to- $k$ hypothesis

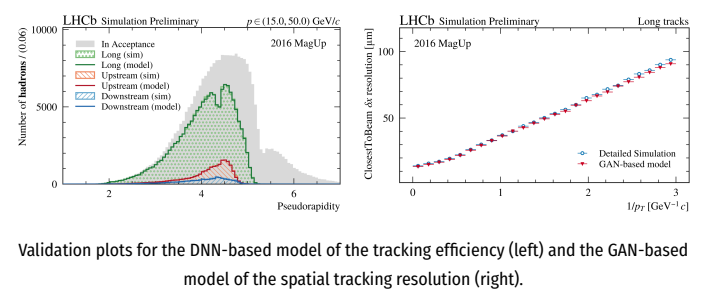
Assuming the existence of an **unambiguous ( $k$ -to- $k$ ) relation** between generated particles and reconstructed objects, the high-level detector response can be modeled in terms of **efficiency** and **"resolution"** (i.e., analysis-level quantities):

- Efficiency:** *Deep Neural Networks* (DNN) trained to perform classification tasks so that they can be used to parameterize the fraction of "good" candidates (e.g., accepted, reconstructed, or selected).
- Resolution:** Conditional *Generative Adversarial Networks* (GAN) trained on detailed simulated samples to parameterize the high-level response of LHCb detector (e.g., reconstruction errors, differential log-likelihoods, or multivariate classifier output).

## 5. Charged particles branch: the tracking system

Lamarr parameterizes the high-level response of the **LHCb tracking system** relying on the following models:

- propagation:** approximates the trajectory of charged particles through the dipole magnetic field  $\rightarrow$  *parametric model*;
- geometrical acceptance:** predicts which of the generated tracks lay within a sensitive area of the detector  $\rightarrow$  *DNN model*;
- tracking efficiency:** predicts which of the generated tracks in the acceptance are properly reconstructed by the detector  $\rightarrow$  *DNN model*;
- tracking resolution:** parameterizes the errors introduced by the reconstruction algorithms to the track parameters  $\rightarrow$  *GAN model*;
- covariance matrix:** parameterizes the uncertainties assessed by the Kalman filter procedure  $\rightarrow$  *GAN model*.

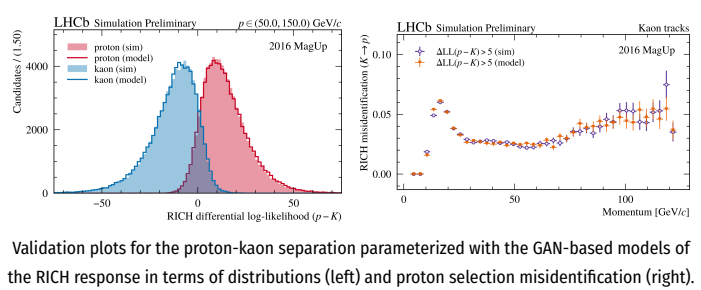


## 6. Charged particles branch: the PID system

Lamarr parameterizes the high-level response of the **LHCb PID system** relying on the following models:

- RICH PID:** parameterizes DLLs resulting from the RICH detectors  $\rightarrow$  *GAN model*;
- MUON PID:** parameterizes likelihoods resulting from the MUON system  $\rightarrow$  *GAN model*;
- isMuon flag:** parameterizes the response of a FPGA-based criterion for muon loose boolean selection  $\rightarrow$  *DNN model*;
- Global PID:** parameterizes the global high-level response of the PID system, consisting of CombDLLs and ProbNNs  $\rightarrow$  *GAN model*.

Lamarr provides separated models for **muons, pions, kaons, and protons** for each PID set of variables.



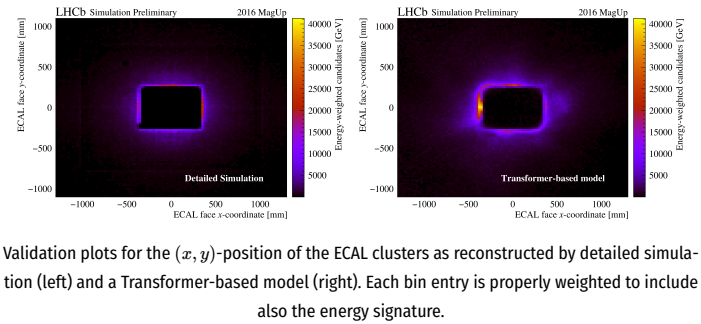
## 7. Neutral particles branch: the ECAL detector

The flash simulation of the LHCb ECAL detector is a non trivial task:

- bremsstrahlung radiation, converted photons, or merged  $\pi^0$  may lead to have  $n$  **generated particles** responsible for  $m$  **reconstructed objects** (in general, with  $n \neq m$ );
- the **particle-to-particle correlation problem** limits the validity of strategies used for modeling the unambiguous  $k$ -to- $k$  detector response.

To parameterize a generic  $n$ -to- $m$  response of the ECAL detector, solutions inspired by the natural language **translation problem** are currently under investigation:

- the aim is to define an **event-level description** of the ECAL response;
- assuming ordered sequences of photons/clusters, the problem can be modeled with a **Transformer** model;
- complying with the problem topology, the ECAL response can be modeled with a **Graph Neural Network** (GNN) model



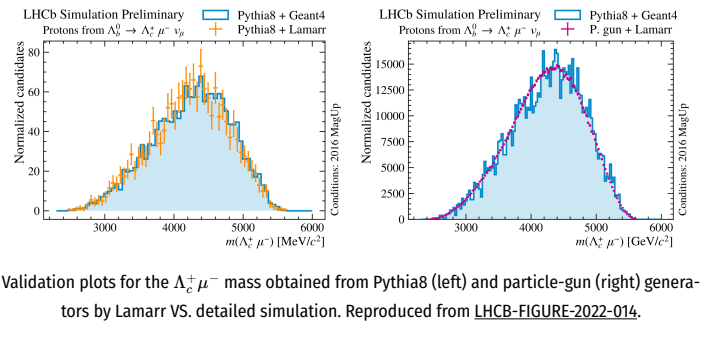
## 8. Validation campaign

Lamarr provides the high-level response of the LHCb detector by relying on a **pipeline of** (subsequent) **ML-based modules**. To validate the charged particles chain, the distributions of a set of **analysis-level** reconstructed quantities resulting from Lamarr have been compared with those obtained from detailed simulation for  $\Lambda_b^0 \rightarrow \Lambda_c^+ \mu^- X$  decays with  $\Lambda_c^+ \rightarrow p K^- \pi^+$ .

The deployment of the ML-based models follows a **transcompilation approach** based on **scikitnC**. The models are translated to C files, compiled as *shared objects*, and then dynamically linked in the LHCb simulation software (Gauss).

The integration of Lamarr with Gauss enables:

- interface with all the **LHCb-tuned physics generators** (e.g., Pythia8, EvtGen);
- compatibility with the **distributed computing middleware** and production environment;
- providing **ready-to-use datasets** for centralized analysis.



## 9. Preliminary timing studies

Overall time needed for producing simulated samples has been analyzed for detailed simulation (Geant4-based) and Lamarr. When Lamarr is employed, the generation of particles from collisions (e.g., with Pythia8) becomes the new **major CPU consumer**.

Lamarr could allow to reduce the CPU cost for the simulation of (at least) **two-order-of-magnitude**. Further reductions will require speeding up the generators.

<b>Detailed simulation:</b> Pythia8 + EvtGen + Geant4 + reco 1M events @ 2.5 kHS06.s/event $\simeq$ 80 HS06.y
<b>Flash simulation:</b> Pythia8 + EvtGen + Lamarr 1M events @ 0.5 kHS06.s/event $\simeq$ 15 HS06.y
<b>Flash simulation:</b> ParticleGun + EvtGen + Lamarr 100M events @ 1 HS06.s/event $\simeq$ 4 HS06.y

## 10. The role of ICSC for Flash Simulation

The **lifecycle of a generic flash-simulation model** includes designing, training, optimization, deployment, and validation, before to be put into production. While the development steps often involve **multiple GPU nodes** (*HPC paradigm*), the validation phase typically relies on the same distributed computing resources employed in the **production environment** (*HTC paradigm*).

The aim of **ICSC** (*Italian Center for SuperComputing*) is to create the national digital infrastructure for research and innovation, leveraging existing **HPC, HTC** and **Big Data** infrastructures and evolving towards a cloud data-lake model. The Lamarr framework is pioneering such **hybrid workloads on distributed and federated resources**, employing nodes from both WLCG data centers and pre-exascale supercomputers (e.g., Leonardo).

## 11. Conclusions and outlook

Great effort is ongoing to put a **fully parametric simulation** of the LHCb experiment into production, aiming to reduce the pressure on computing resources.

DNN-based and GAN-based models succeed in describing the high-level response of the LHCb tracking and PID detectors for **charged particles**. Work is still required to parameterize the response of the ECAL detector due to the **particle-to-particle correlation problem**.

Future development Lamarr aims to support both integration within the LHCb software stack and its use as a **stand-alone** package.

## Acknowledgements

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