



Particle Identification with Machine Learning for ALICE ITS2

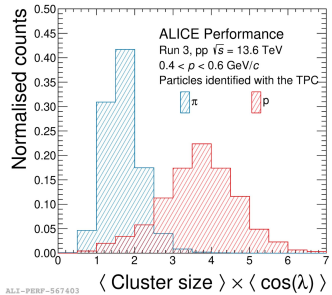
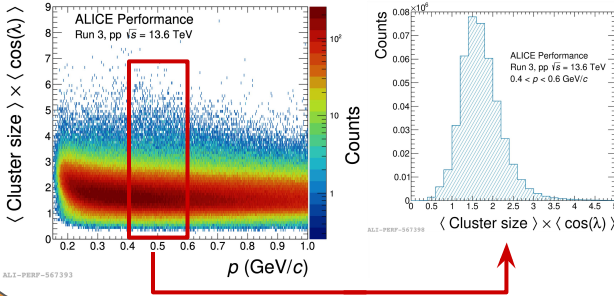


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- Particle identification (PID) information is provided by various ALICE detectors through complementary experimental techniques
- The Run 2 ALICE Inner Tracking System (ITS1) provided PID information for low-momentum particles by measuring their specific energy loss
- Upgraded ITS for the LHC Run 3 (ITS2)¹:
 - higher granularity (pixel size reduced from 50 μm × 425 μm to 29.24 μm × 26.88 μm) → better spatial resolution (5 μm)
 - lower material budget (0.36 X₀ %)
 - it can cope with higher interaction rates (50 kHz in Pb–Pb collisions)
 - digital readout → ITS2 cannot measure the energy loss directly

- Cluster size, i.e. the number of pixels turned on by a charged particle traversing a layer of the ITS2, used as a proxy of the particle energy loss

- cluster size corrected by particle's incident angle on the ITS layer (λ)
- cluster size distribution for different species is not separated



- Particle distributions normalised per species
 - π and p tracks tagged with the ALICE Time-Projection Chamber (TPC)
- Sensitivity to the different energy loss in the detector material

Particle identification with Machine Learning

- Separation can be improved with Machine Learning (ML)
- Boosted Decision Tree (BDT) employed for regression
 - inference of the particle $\beta \rightarrow \beta_{\text{BDT}}$
- Training sample:
 - ITS-TPC matched tracks, particles tagged with the TPC information
 - full data-driven approach
 - training variables: cluster size (average and on different ITS2 layers), momentum (p), $\sin(\varphi)$ (φ is the particle angle on the transverse plane), $\tan(\lambda)$
- Regressor implemented by XGBoost¹
 - hyperparameters of the BDT optimised with Optuna²

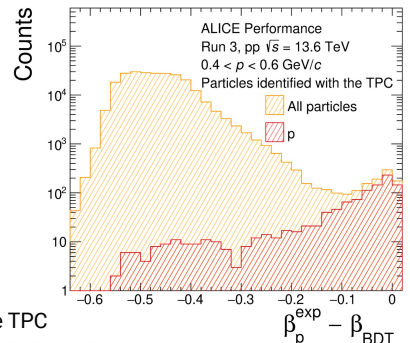
$$\beta = \frac{p}{\sqrt{p^2 + m^2}}$$

dmlc
XGBoost



Results and outlook

- Regressor trained and applied on Run 3 data
 - pp collisions at $\sqrt{s} = 13.6$ TeV
- BDT performance evaluated for protons with $0.4 < p < 0.6$ GeV/c through the discrepancy between β_{BDT} and β^{exp}
- Proton distribution peaked around zero
 - well separated from the rest of the particles
 - PID at low momentum feasible with ML
- ITS PID will be used for:
 - identifying low-momentum particles
 - improving the matching with TPC
 - analysis of (hyper)nuclei decayed or absorbed before reaching the TPC



¹ ALICE Collaboration, J.Phys. G 41 (2014) 087002

² Chen et al., 22nd SIGKDD Conference on Knowledge Discovery and Data Mining, 2016

³ Takuya A. et al, KDD (2019)