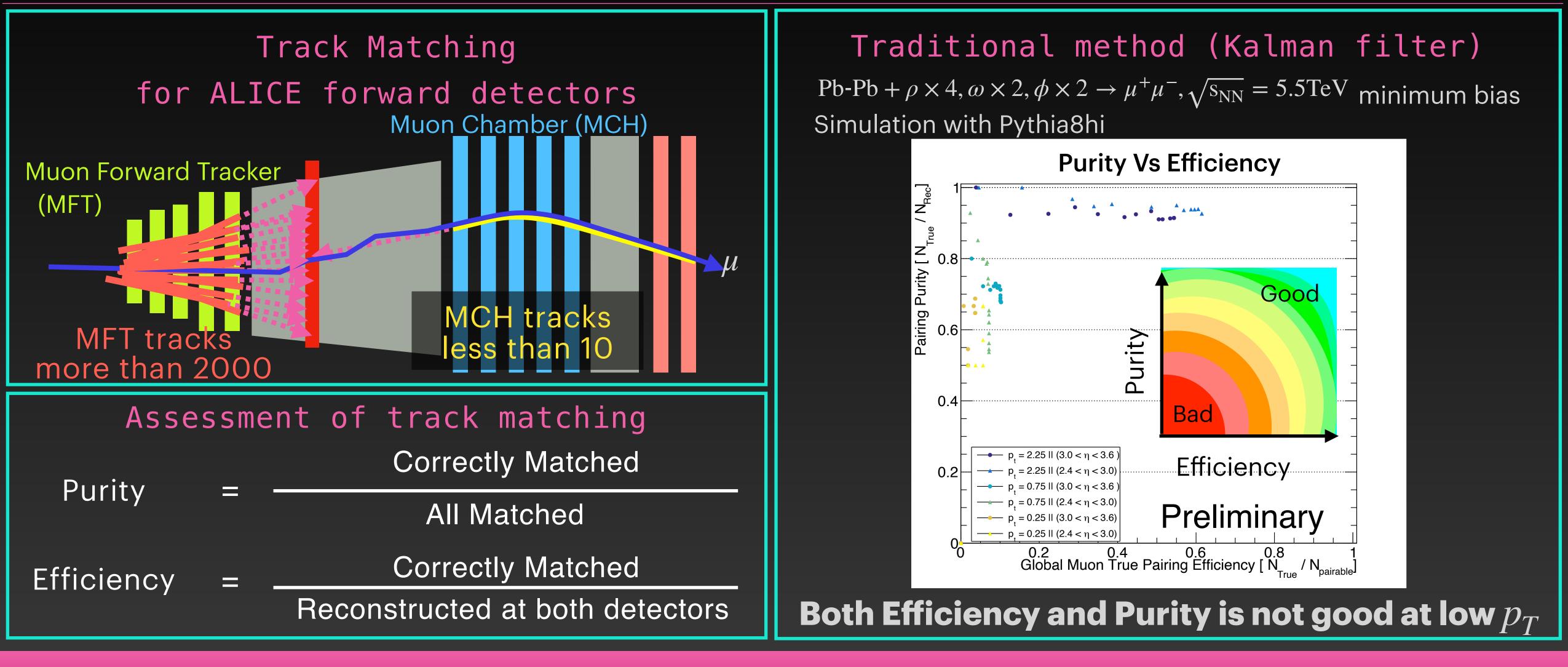
Performance Evaluation of Forward Muon Track Matching in ALICE Run 3



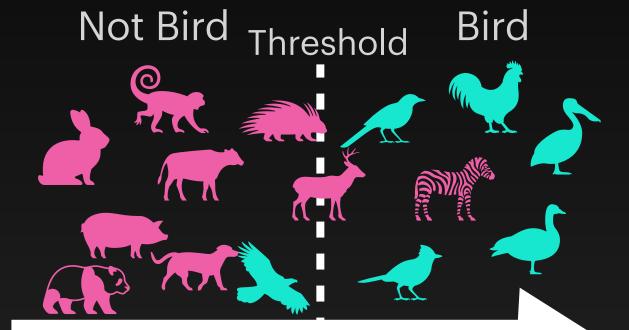
Issue : Traditional method's result is not enough...

Ren Ejima for the ALICE collaboration, Hiroshima Univ. Japan

Track Matching with Machine Learning

Classification by machine learning

Example: Classify whether Bird or any other animals Bird



Predicted probability of being bird

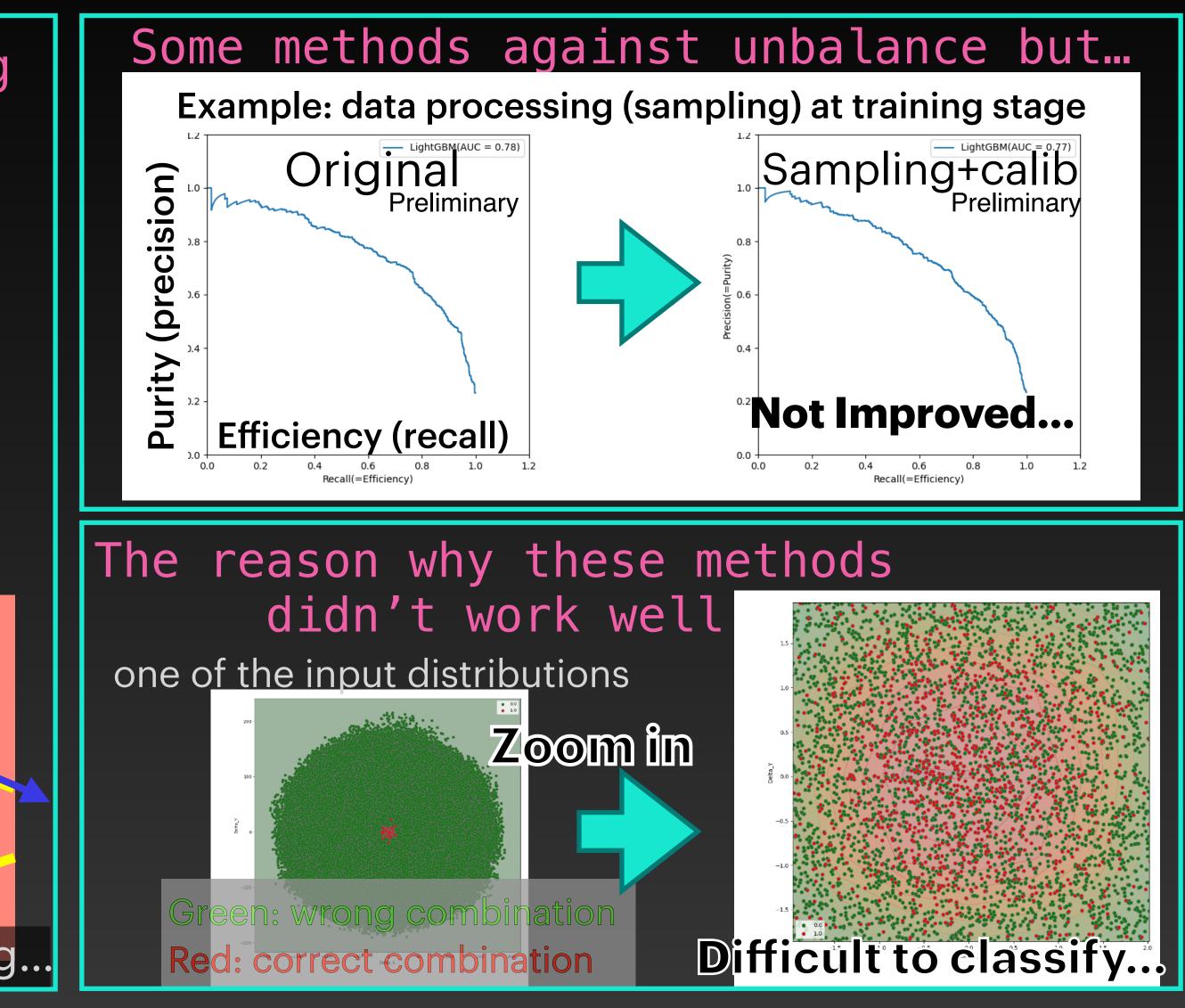
Learn to improve purity and efficiency of classification.

Track Matching: Classify whether Correct or Wrong

MFT tracks more than 2000

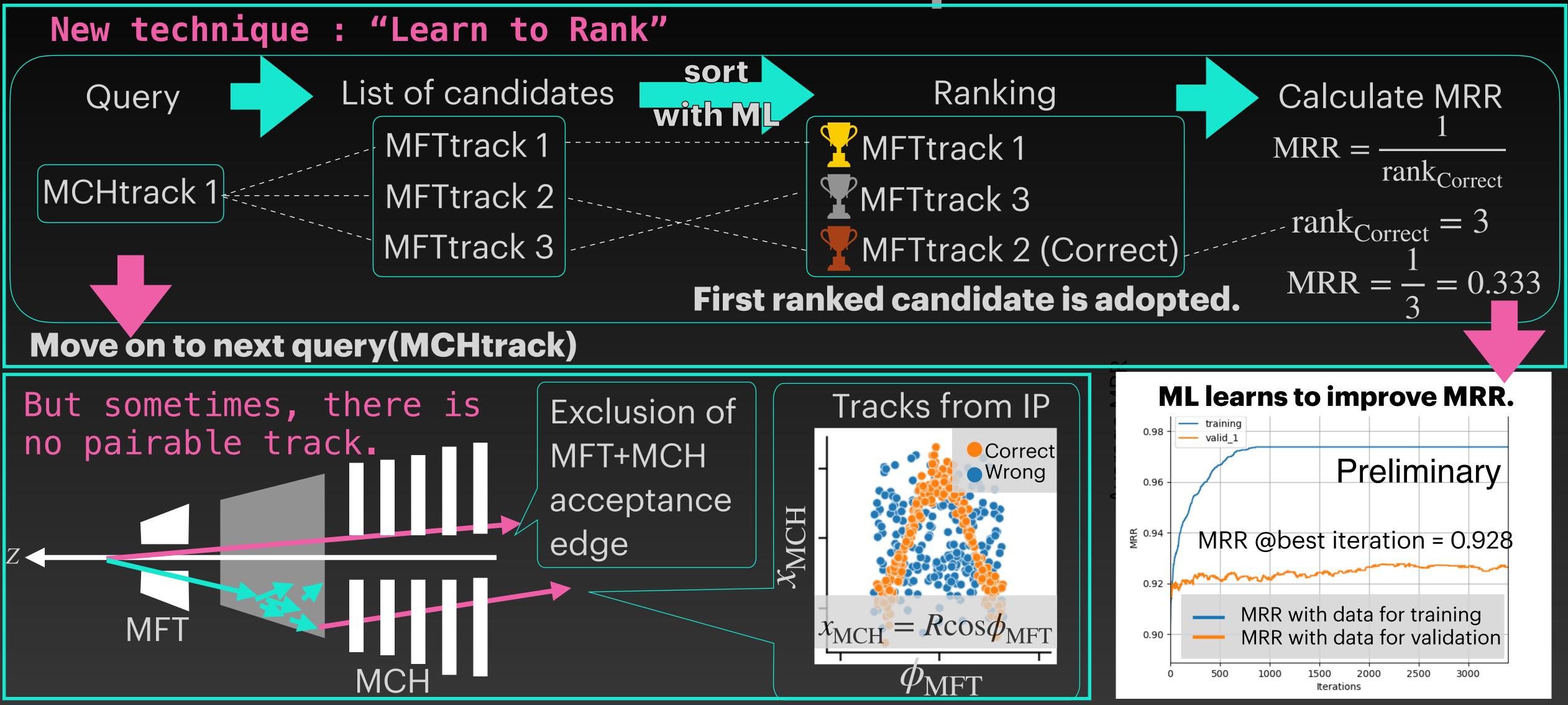
Unbalance (Correct << Wrong) will disturb training...

than



Difficult to classify ...





Idea : Learn of each guery and learn to improve ranking

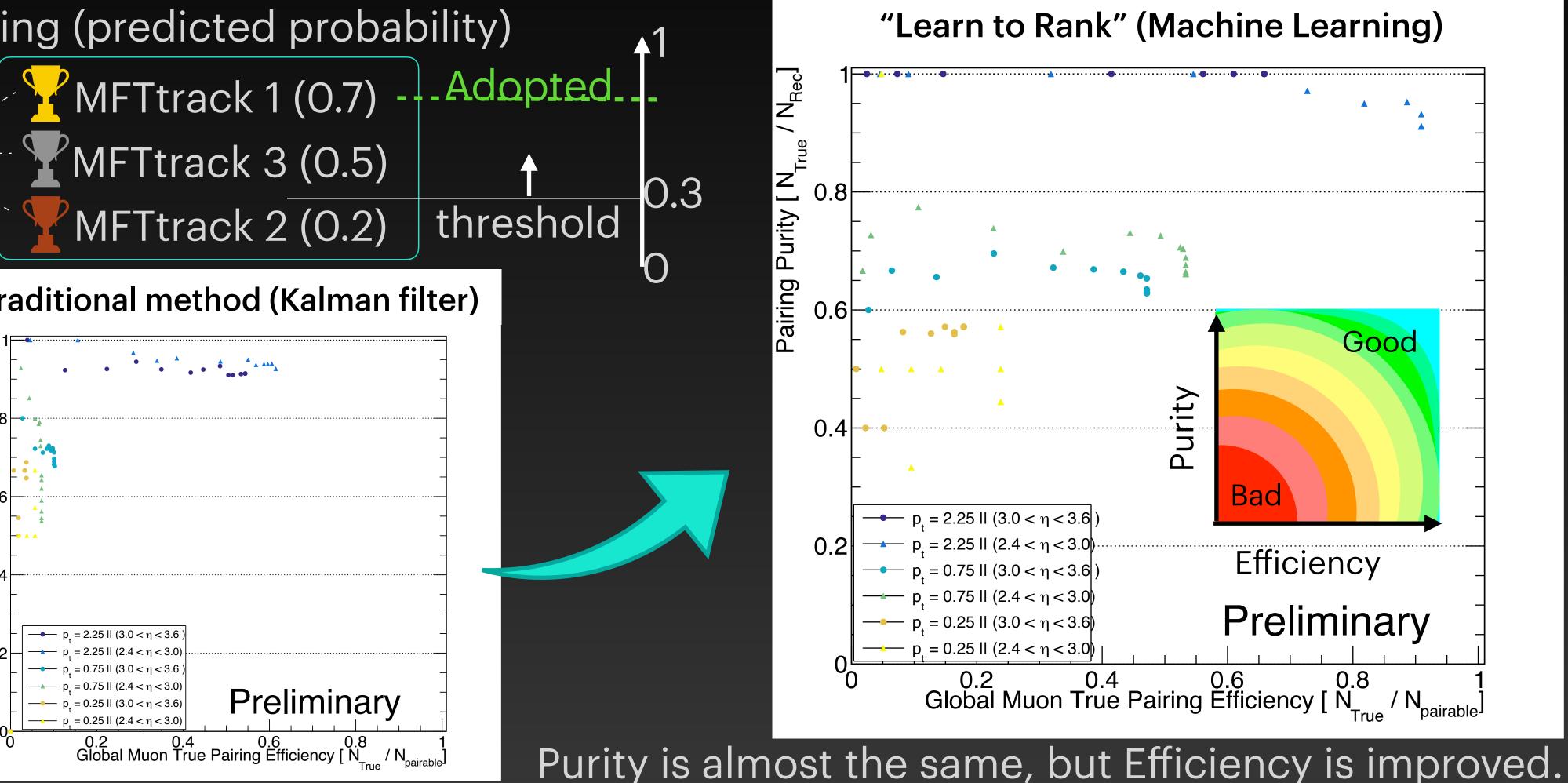
New Technique

Result of Machine Learning

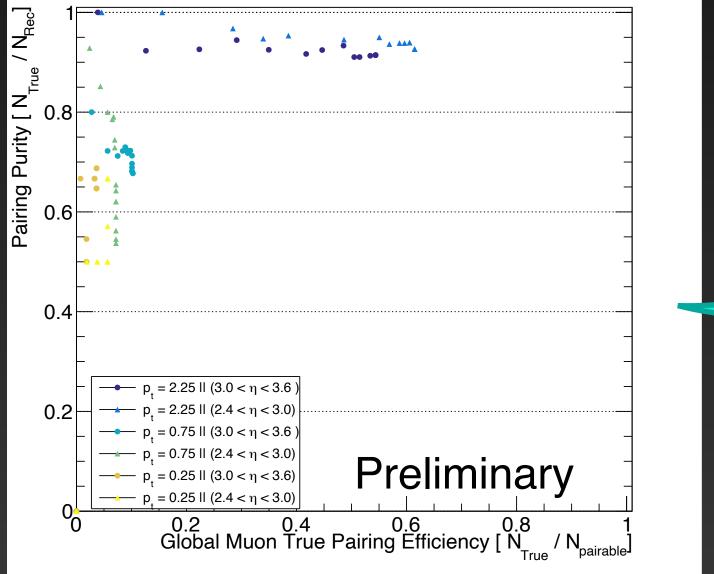
Purity Vs Efficiency for different thresholds of predicted probability

Ranking (predicted probability)

Query MCHtrack



Traditional method (Kalman filter)



Result : Improved



Summary and Outlook

Summary

- Track Matching for ALICE forward detectors is very challenging due to high multiplicity and multiple scattering
- improve results due to overlapping of the distribution of feature parameters
- Data processing at training stage and weighted learning didn't contribute to •New technique "Learn to Rank" helps to improve result.

Outlook

Reduce the number of candidates with the highest prediction



MFTtrack 1 (0.7) MFTtrack 3 (0.7) MFTtrack 2 (0.2)

Thank you !

