## Jet momentum reconstruction in the QGP background with machine learning

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# Outline

### Simulation Models: PYTHIA & LBT models

### Background Subtraction Methods

Set Momentum Reconstruction

✤ Jet Nuclear Modification Factor R<sub>AA</sub>



## **QGP** and Jet Modifications



- Quark-gluon plasma (QGP) in heavy-ion collisions: deconfined phase, hot dense medium.
- hard probe to medium properties.
- measure the quenching effects.



- Vacuum jets: PYTHIA8 model
- Jet interaction with QGP: Linear Boltzmann Transport (LBT) model
- \* Pb-Pb collisions in 0-10% centrality at  $\sqrt{S=5.02TeV}$
- QGP background: a toy thermal model
- \* Reconstructed jets with anti- $k_T$ , R=0.4
- Target  $p_T$  sum of the (PYTHIA/LBT) jet particle  $p_T$ within jet cone, in the presence of background particles



# **Background Subtraction Methods**

### Area-based method

- Event-by-event basis: background momentum density  $\rho$
- For each jet: reconstructed jet momentum  $p_T^{\text{rec}} = p_T^{\text{raw}} - \rho A$
- Leading to large residual fluctuations
- Constituent Subtraction method
  - Local subtraction of soft background
  - Simultaneously correcting the 4-momentum of the jet and its substructures

ML techniques with following PYTHIA jet observables as inputs

The uncorrected jet momentum

- The jet transverse momentum, corrected by the area-based method
- Jet mass, radial moment, momentum dispersion, and LeSub
- The number of constituents within the jet
- Mean and median of all constituent transverse momenta
- The transverse momenta of the first ten leading



R. Haake and C. Loizides, PHYS. REV. C 99, 064904 (2019)



### Jet Momentum Reconstruction in ALICE



<sup> $\diamond$ </sup> More precise estimate for low  $p_T$  jet

\* PYTHIA & Quenching-aware Variants are employed to train DNNs separately to predict jet  $p_T$  and estimate  $R_{AA}$ 

How about training the datasets together?



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## **Jet Momentum Reconstruction**



Test PYTHIA-trained ML on different datasets

PYTHIA-trained ML has a prediction bias when applied on LBT jets.

The bias is reduced when the ML model is applied on LBT jets without recoil particles: Recoil particles may be considered as the background.



## Jet Momentum Reconstruction



### Testing on LBT data

- The bias is reduced when the ML model is trained by LBT jets directly or by LBT+PYTHIA jets.
- ML methods are more accurate than conventional methods (Area-based & Constituent Subtraction).

### **Testing on PYTHIA data**

\* LBT+PYTHIA-trained ML can make accurate predictions on PYTHIA and LBT data simultaneously, demonstrating strong robustness.





\*  $R_{AA}$  from ML(LBT) & ML(LBT+PYTHIA) are closer to that of the current target baseline "LBT jet within bkg." than  $R_{AA}$  from ML(PYTHIA).





jets reconstructed within background (LBT jet within bkg.) VS jets reconstructed without background (LBT-only jets).

Significant discrepancy between two baselines, i.e., Experiment vs Theory:





 We apply a matching procedure to reveal the contribution from fake jets and the p<sub>T</sub> difference between matched jets.



## Jet Momentum Reconstruction



LBT-trained ML with Matching Procedure

\*  $p_T$  of LBT jets within bkg. are generally larger than  $p_T$  of LBT-only jets, leading to an over-estimation of  $R_{AA}$ 

\* ML models are dedicated to predict LBT-only jet  $p_T$  from LBT jet within bkg.







 $R_{AA}$  from LBT-only jet can be well reproduced with such ML setups

\* By incorporating fake jets and labelling their  $p_T = 0$  in the training, the performance will improve



# Summary & Outlook

- quenched jets simultaneously to obtain strong robustness and generalizability.
- dedicated to predict the  $p_T$  of real target jets in theory from jets reconstructed in experiment.
- Incorporate quenched jets from various MC models.
- Include the realistic anisotropic background and medium response.

To better reconstruct jet momentum with machine leaning, we suggest training the ML model with various

\* To more accurately obtain the jet nuclear modification factor  $R_{AA}$  in theory, we apply a matching procedure and reveal that the contribution from fake jets is relatively small though non-negligible. ML models are

Develop novel ML models assisted with better jet representation to improve the prediction accuracy.

