

Observing jet quenching using generalized jet angularities in Au+Au collisions at $\sqrt{s_{\rm NN}} = 200$ GeV from STAR



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Abstract

Jets originating from hard-scattered partons in the early stages of heavy-ion collisions travel through the Quark Gluon Plasma (QGP) and are modified or quenched relative to a p+p collision baseline. Moments of the jet's transverse momentum (p_{T}) profile in the $\eta - \phi$ plane relative to the jet-axis, called generalized jet angularities are an important class of jet substructure observables to study in medium modifications of the jet's radiation and fragmentation patterns. Previous measurements of these angularities have been performed using quenched jets from Pb+Pb collisions at $\sqrt{s_{NN}} = 2.76$ TeV in the LHC, and similar measurements using heavy-ion collisions at RHIC energies will probe jet quenching in a region of phase space that is complementary to the region probed in the LHC. In this study, we present relative central-preipheral nuclear modification factors ($R_{\rm CP}$) using fully corrected measurements of various generalized jet angularities using jets from Au+Au collisions at $\sqrt{s_{\rm NN}} = 200$ GeV collected by the STAR experiment. We present proof of concept studies done with jets from JEWEL simulation, of a novel machine-learning based method that measures the degree to which quenched and unquenched jets are distinguishable. Measurements are differential in centralities of the Au+Au collisions.



Classifying medium vs vacuum jets

proof of concept study:

Previous results

Previous measurements of Girth, p_T^D in Pb+Pb collisions at $\sqrt{s_{\rm NN}} = 2.76$ TeV at the LHC [1] qualitative agreement with a more quark-like jet fragmentation (higher p_T^D , lower Girth) compared to PYTHIA-6/8)



Can we tell apart unquenched and quenched jets using classification algorithms?



- PYTHIA6 [4] \rightarrow unquenched jets, JEWEL [5] (with/without recoils) \rightarrow quenched jets • Dense neural networks (DNNs) classify jets as quenched or unquenched, trained on 80% of generated jets, Area under ROC curve (AUC) score from remaining 20%, similar to [3]
- Jets \rightarrow 9-D vectors of features: $p_{T,jet}$, $N_{constit.,jet}^{charged}$, $N_{constit.,jet}^{neutral}$, LeSub, λ_0^2 , $\lambda_1^{0.5}$, λ_1^1 , λ_1^2
- Exagerrated ROC curves from JEWEL with recoils due to "background" of recoil centers
- Quenched jets from JEWEL (w.o. recoils) distinguishable from unquenched jets (AUC > 0.5)

AUC score decreases from central (0-10%) to peripheral (50-80%) JEWEL (w.o. recoils) events \implies more modification in central compared to peripheral

Which features contribute most to classification?



- Random forest (RF) of boosted decision trees $[2] \rightarrow$ access feature importance scores
- Quenched jets from JEWEL without recoils
- Feature importances similar across all centrality ranges
- Further analysis to understand relation of importance scores to quenching

DNN vs RF				
Slightly better AUC scores of DNN				
over RF				
	Centrality	RF	DNN	
	0-10%	0.59	0.6	
	10-20%	0.54	0.58	
	20-50%	0.53	0.57	
	50-80%	0.51	0.52	
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Outlook

- Further analysis of $R_{\rm CP}$ to reduce systematic uncertainties
- R_{AA} measurements of generalized angularities using jets from STAR p+p collisions at $\sqrt{s_{\rm NN}} = 200 \; {\rm GeV}$
- Improve classification algorithms to distinguish between quenched and unquenched jets
- one tower with $E_{\text{tower}} \ge 4$ GeV) to enhance
- systematic uncertainties between central,

Further analysis ongoing to improve systematic uncertaities

• Apply classifiers to distinguish between jets from STAR p+p and Au+Au collisions at $\sqrt{s_{\rm NN}} = 200 \; {\rm GeV}$

References

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