

# Probing the QGP with inclusive jets and their substructure

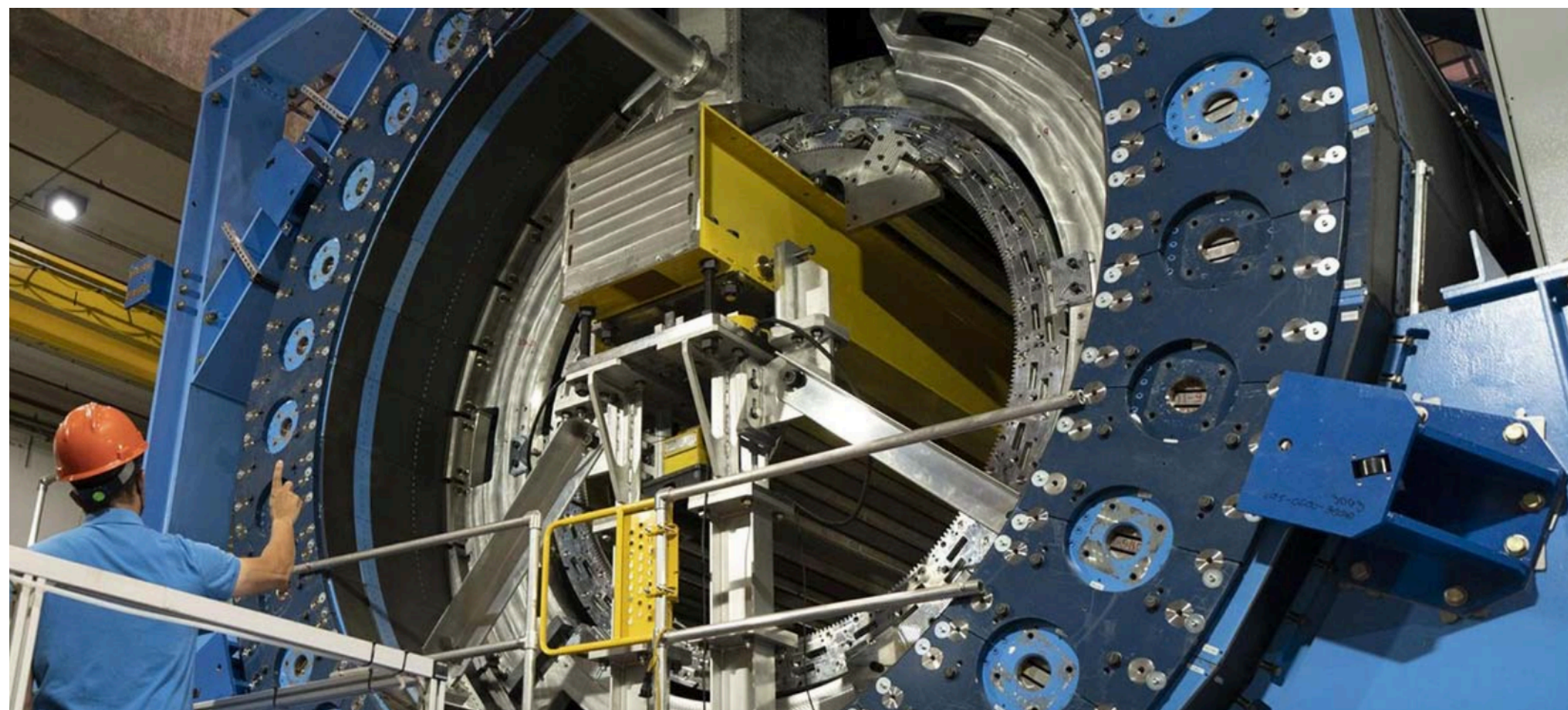
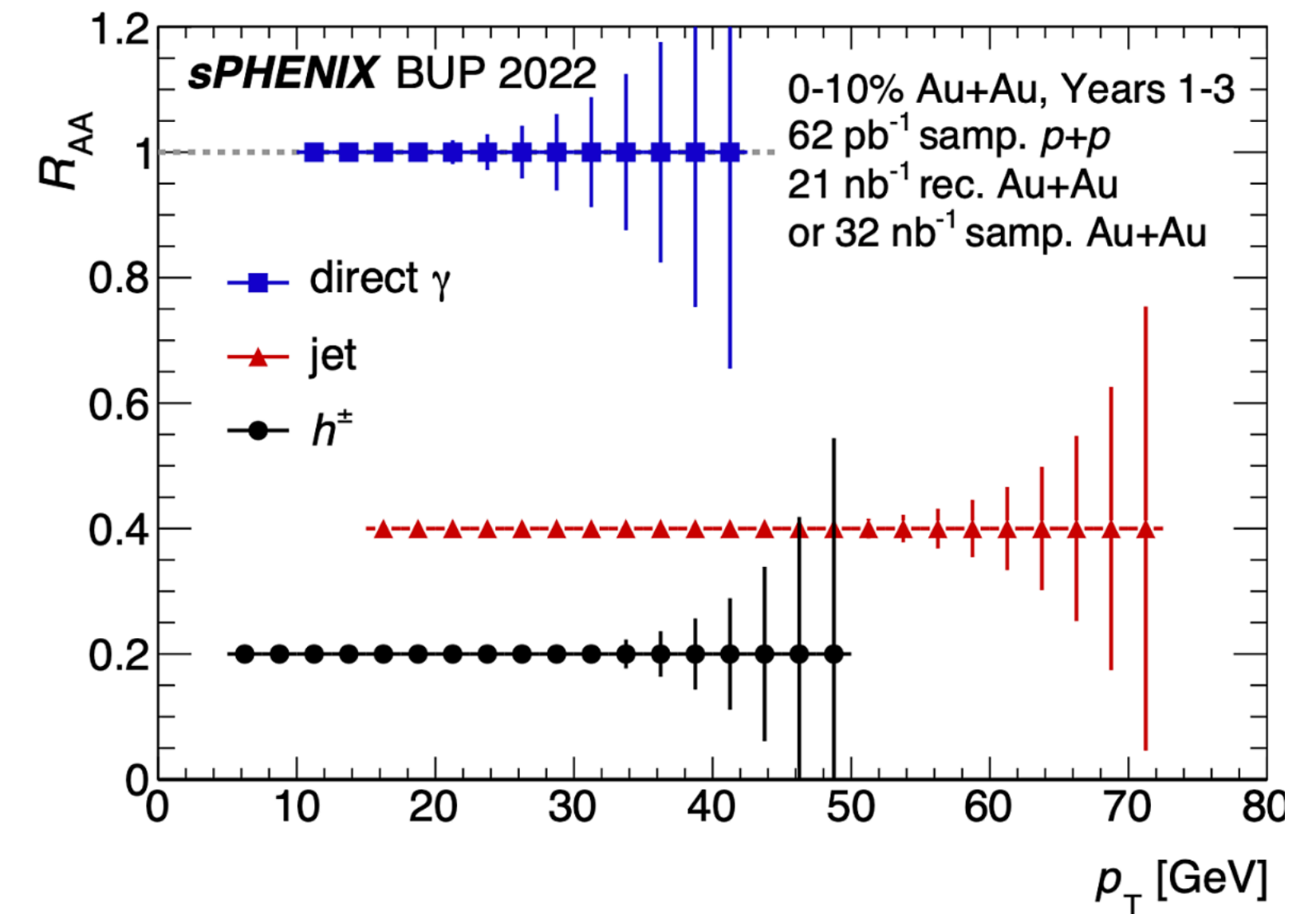
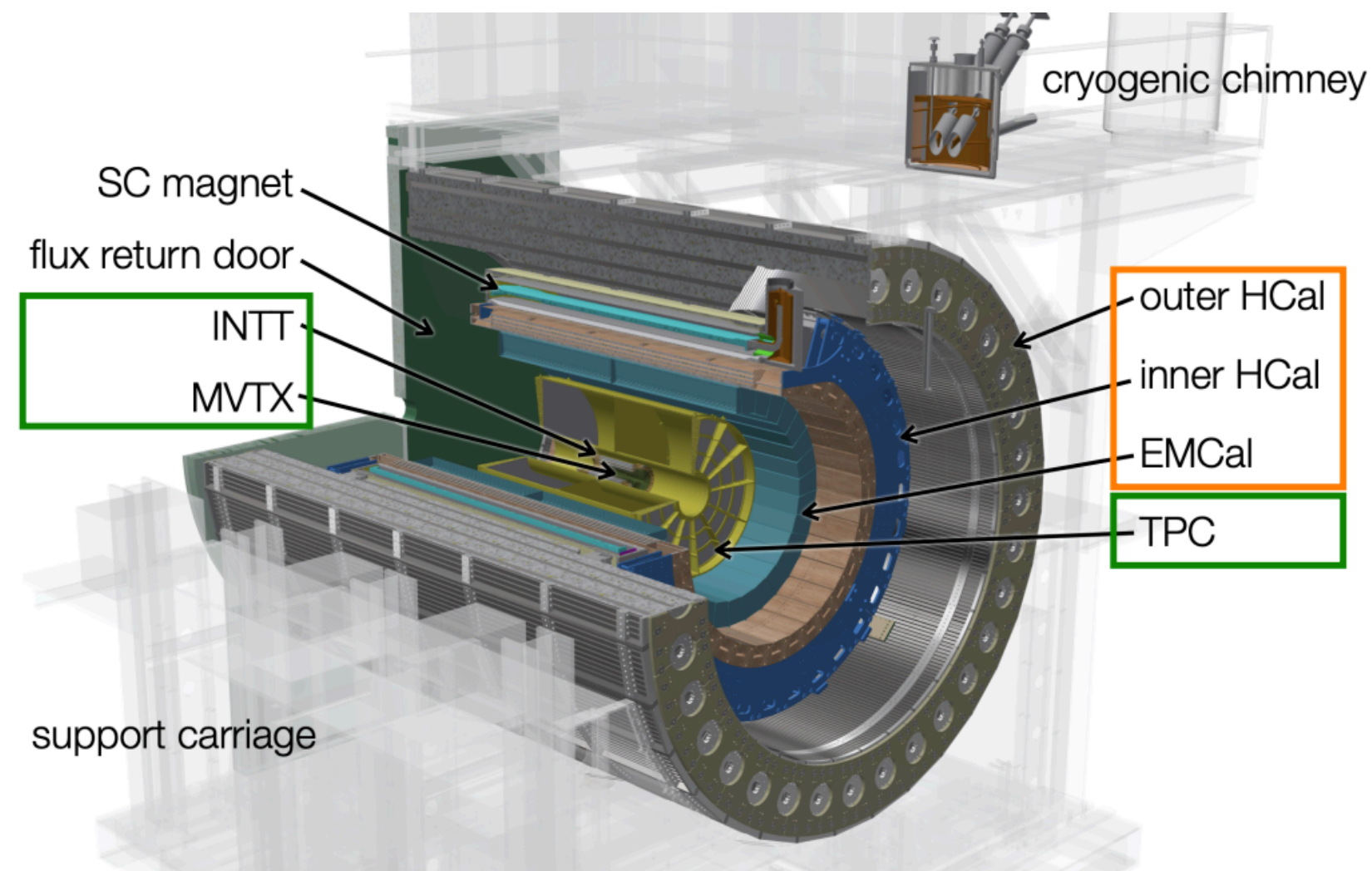
Felix Ringer

Old Dominion University,  
Jefferson Lab

Predictions for sPHENIX, RBRC BNL, 07/21/22



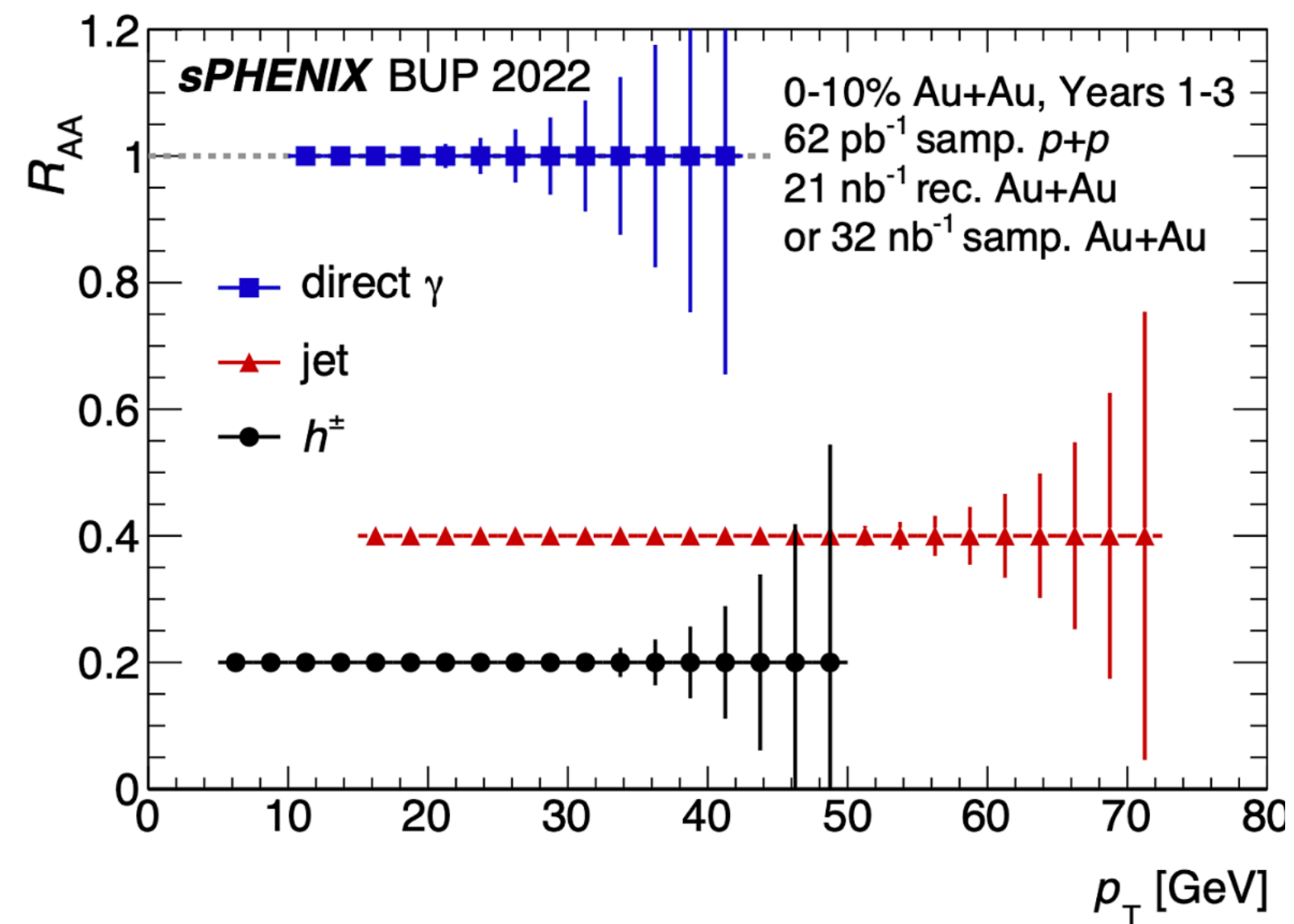
# Jet quenching and energy loss



Precision tests of jet quenching with sPHENIX

# Jet quenching and energy loss

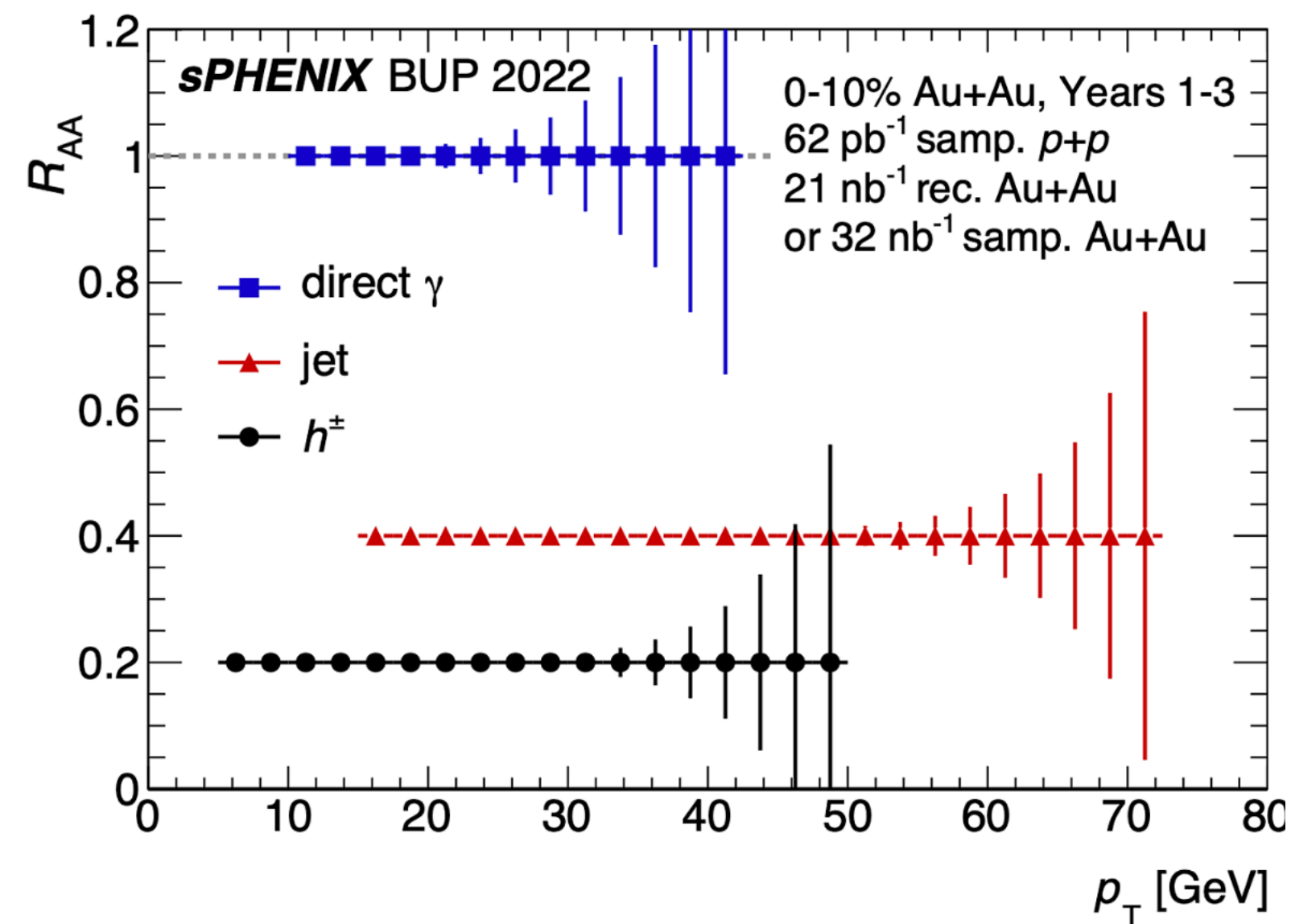
- QCD factorization & universality?



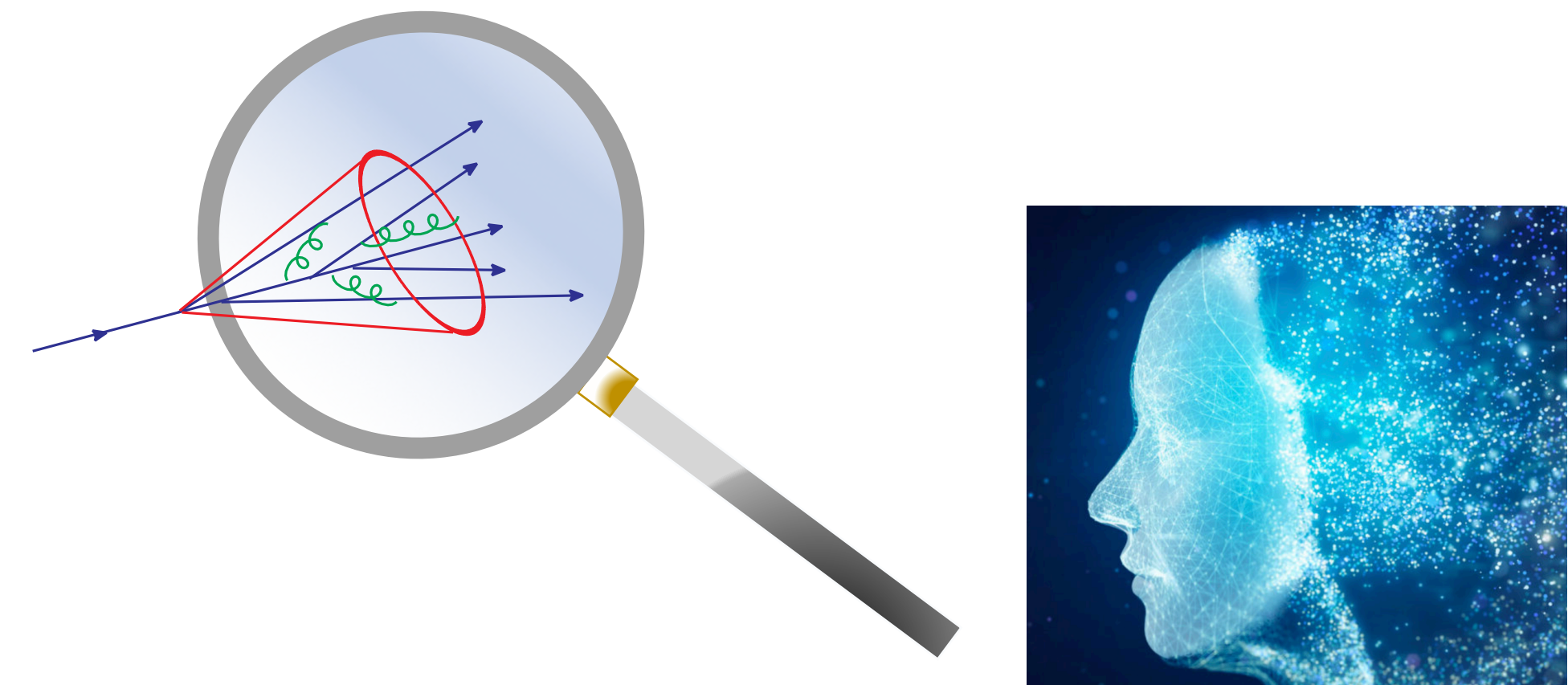
$$\frac{d\sigma_{pp \rightarrow \text{jet} + X}}{d\eta dp_T} = \sum_{ijk} f_{i/p} \otimes f_{j/p} \otimes H_{ijk} \otimes J_k^{\text{med}}$$

# Jet quenching and energy loss

- QCD factorization & universality?



- What is the (maximal) information content of jet substructure observables?
- How many jet substructure observables do we need to measure?
- How to systematically make use of all the information provided by sPHENIX?



$$\frac{d\sigma_{pp \rightarrow \text{jet} + X}}{d\eta dp_T} = \sum_{ijk} f_{i/p} \otimes f_{j/p} \otimes H_{ijk} \otimes J_k^{\text{med}}$$

# Outline

Introduction

Inclusive jets &  
quark/gluon fractions

Machine-learned  
jet substructure  
observables

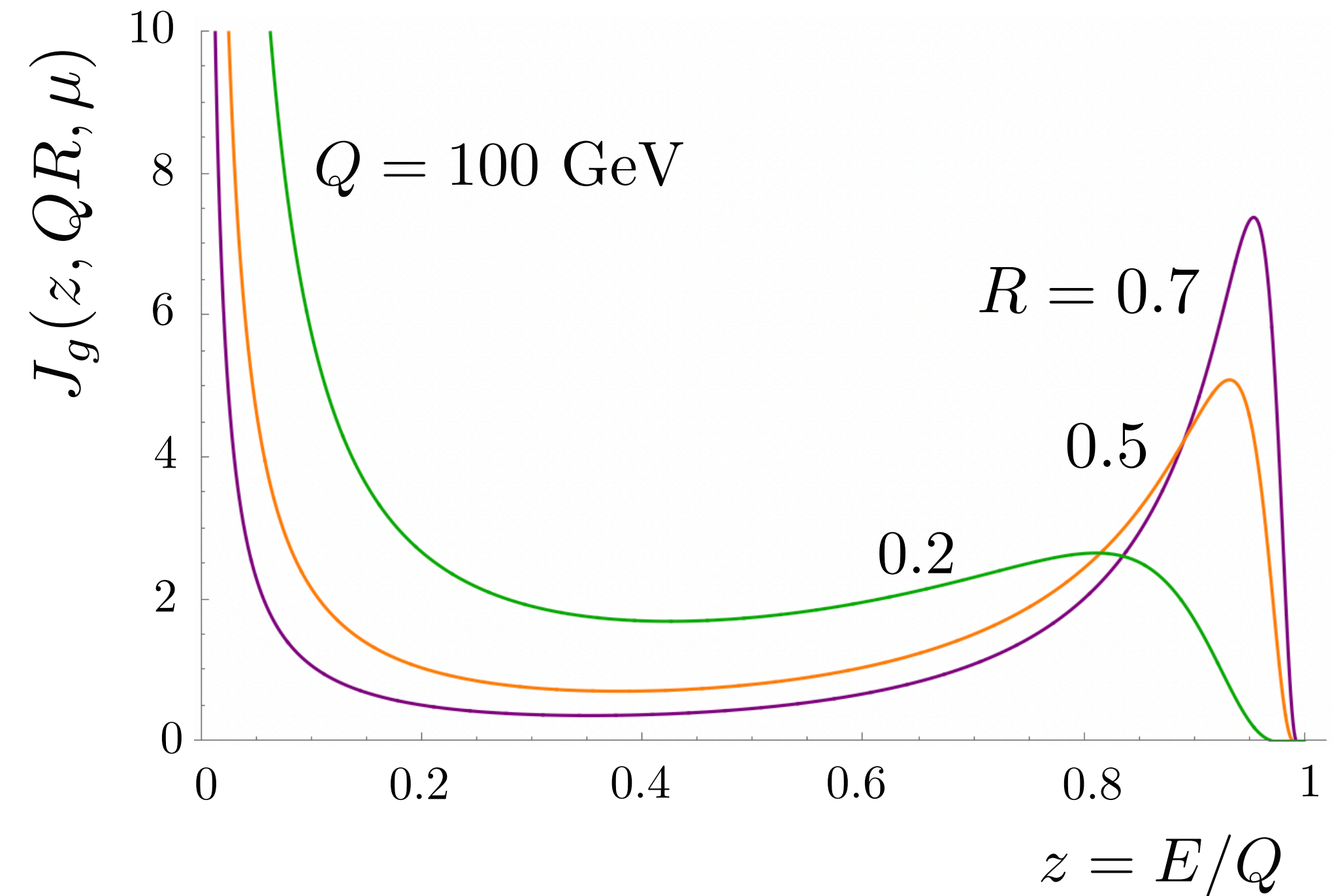
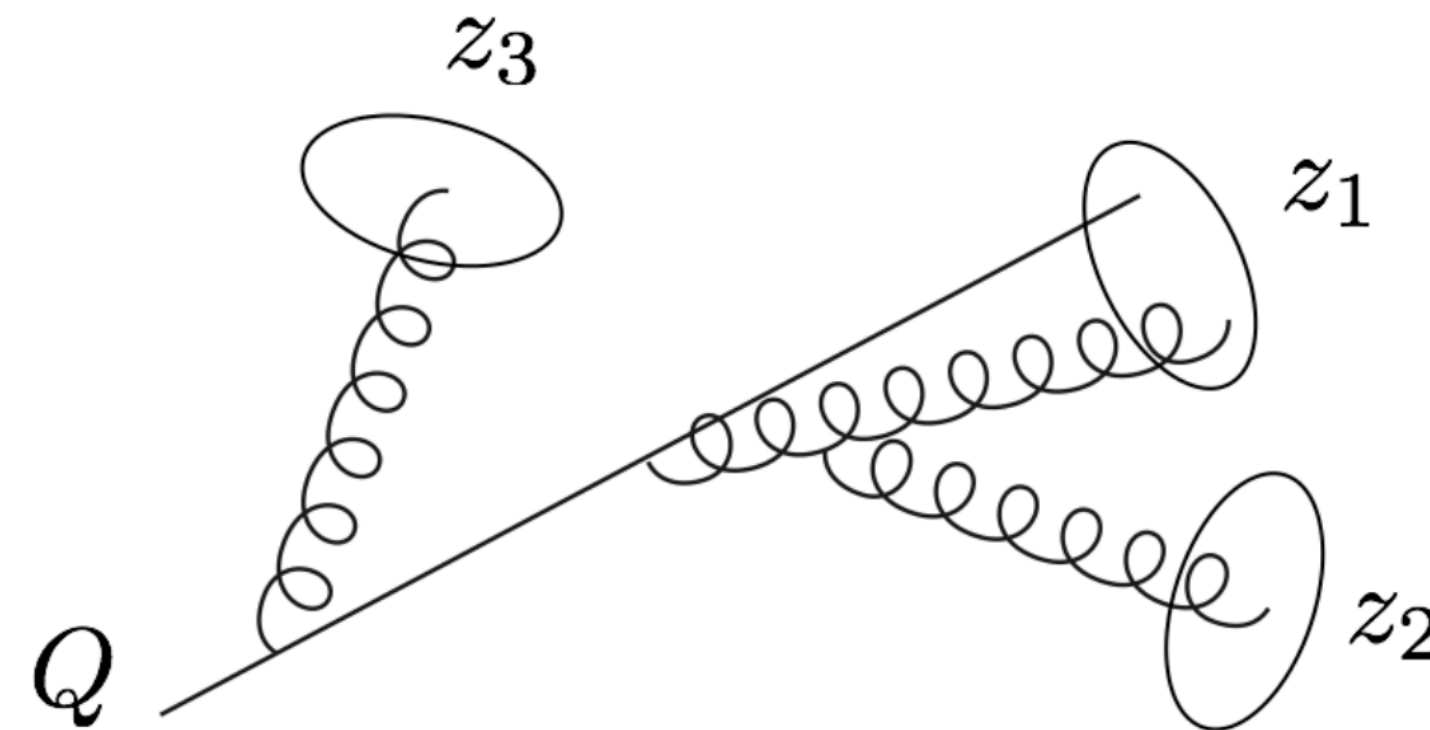
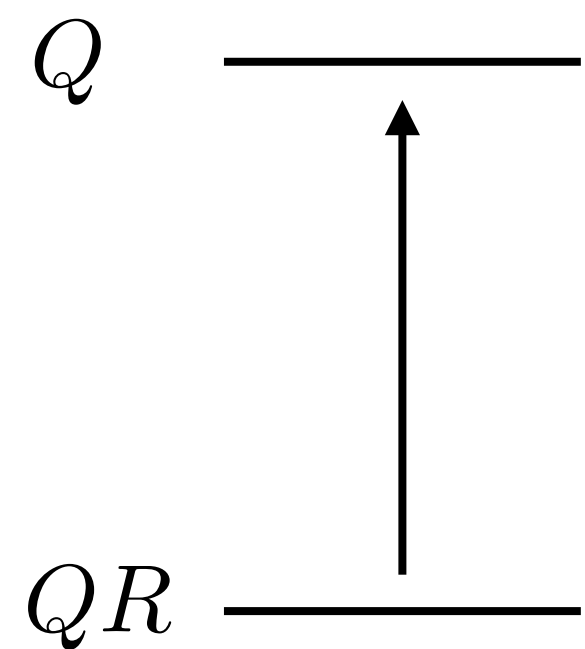
# Inclusive jet cross sections

- **Factorization**

$$\frac{d\sigma_{pp \rightarrow \text{jet} + X}}{d\eta dp_T} = \sum_{ijk} f_{i/p} \otimes f_{j/p} \otimes H_{ijk} \otimes J_k$$

- **DGLAP evolution**

$$\mu \frac{d}{d\mu} J_i = \frac{\alpha_s}{2\pi} \sum_j P_{ji} \otimes J_j$$



Dasgupta, Dreyer, Salam, Soyez `14  
 Kaufmann, Mukherjee, Vogelsang `15  
 Kang, Ringer, Vitev `16  
 Dai, Kim, Leibovich `16  
 Liu, Moch, Ringer `18, `19

# Inclusive jets in heavy-ion collisions

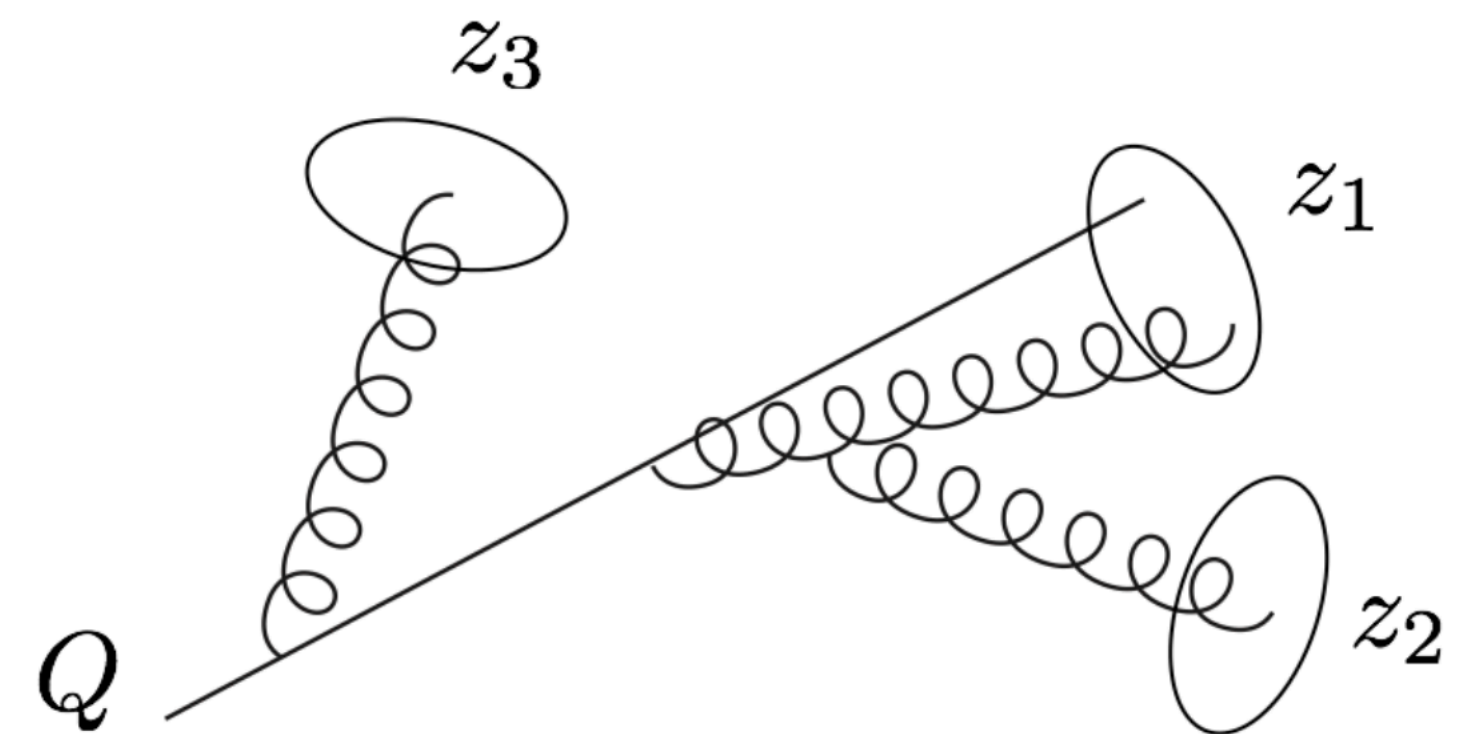
- **Factorization** - can we systematically extend vacuum factorization theorems in vacuum to heavy-ion collisions?

$$\frac{d\sigma_{pp \rightarrow \text{jet} + X}}{d\eta dp_T} = \sum_{ijk} f_{i/p} \otimes f_{j/p} \otimes H_{ijk} \otimes J_k \quad \longrightarrow$$

Heavy-ion collisions?  $J_k^{\text{med}}$

- **Quark/gluon fractions** for jet substructure observables

$$f_q J_q(\tau) + f_g J_g(\tau)$$



# Inclusive jets in heavy-ion collisions

Qiu, FR, Sato, Zurita '19

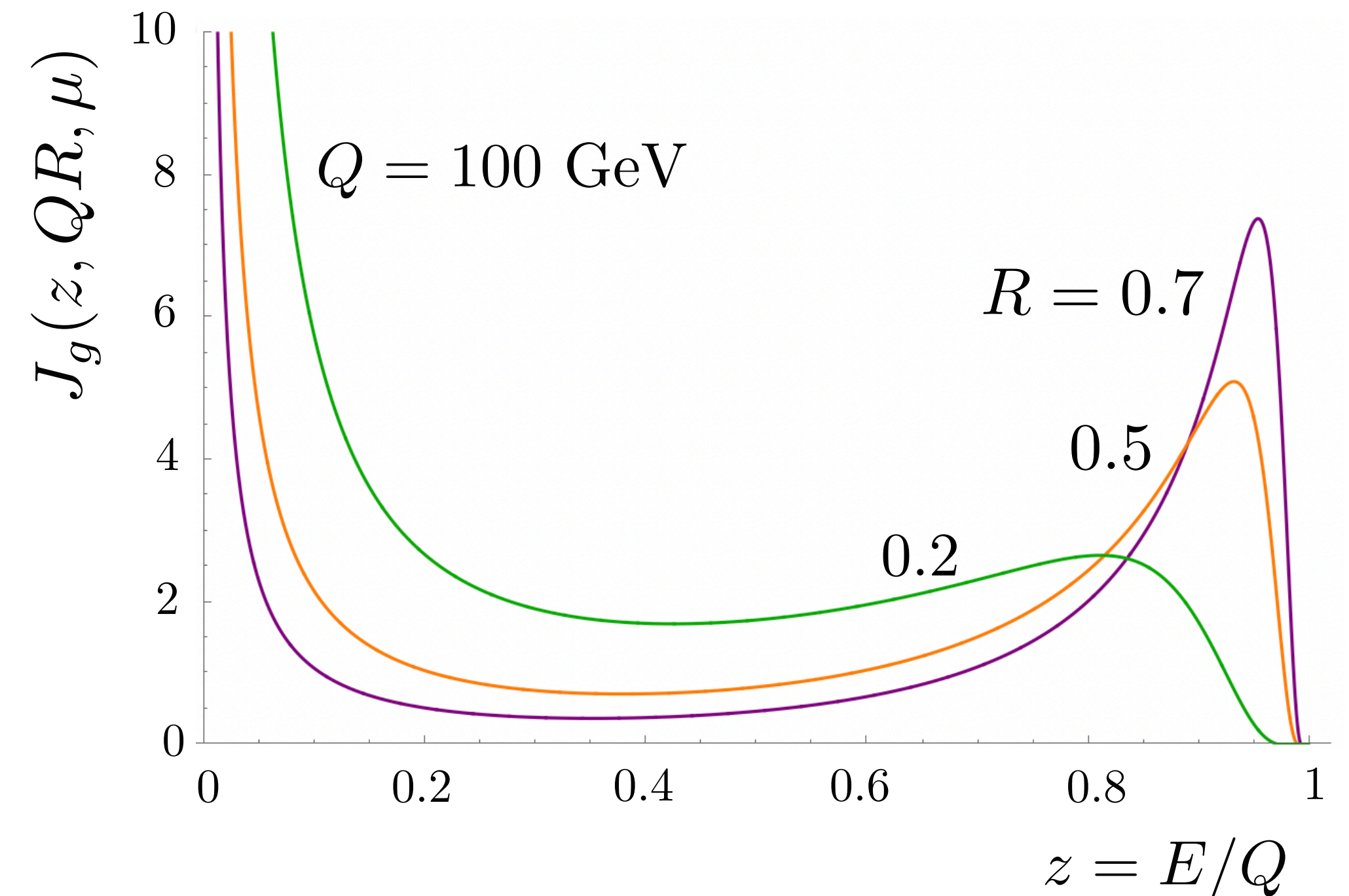
- Introduce medium modified jet function at the jet scale

$$J_c^{\text{med}}(z, p_T R, \mu_J) = W_c(z) \otimes J_c(z, p_T R, \mu_J)$$

$$W_c(z) = \epsilon_c \delta(1 - z) + N_c z^{\alpha_c} (1 - z)^{\beta_c}$$

- Monte Carlo sampling approach

NNPDF '17, JAM '16



nPDFs

Eskola, Paakkinen, Paukkunen, Salgado '17, Kovarik et al. '16  
de Florian, Sassot, Zurita, Stratmann '12

nFFs

Sassot, Stratmann, Zurita '10

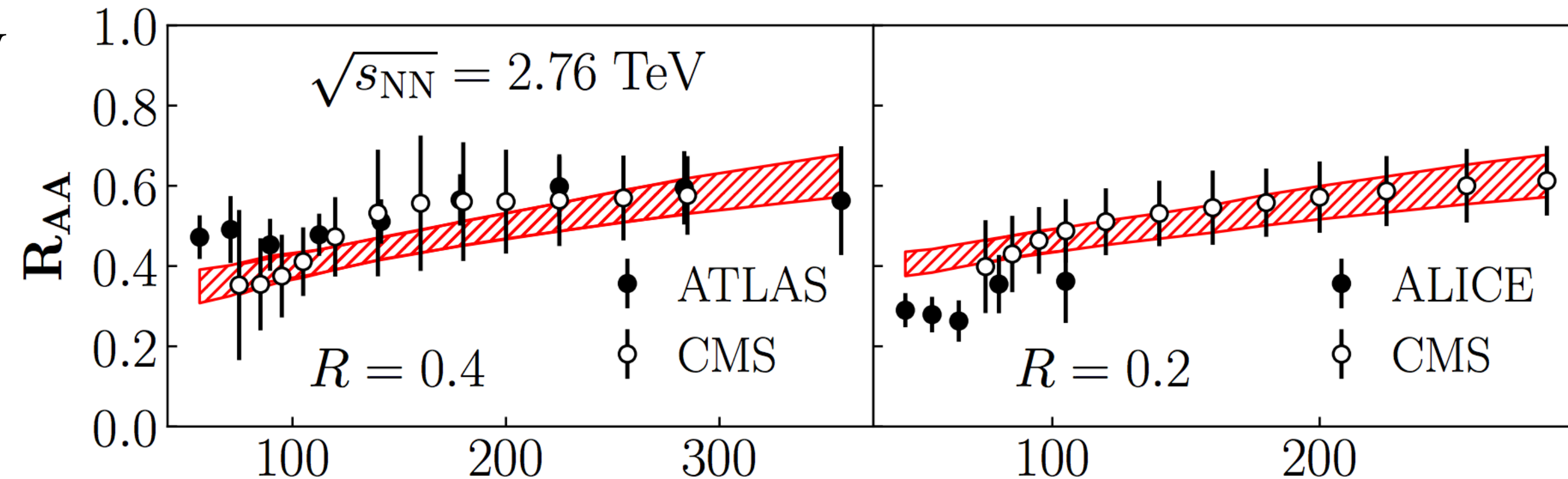


# Inclusive jets in heavy-ion collisions

Qiu, FR, Sato, Zurita '19

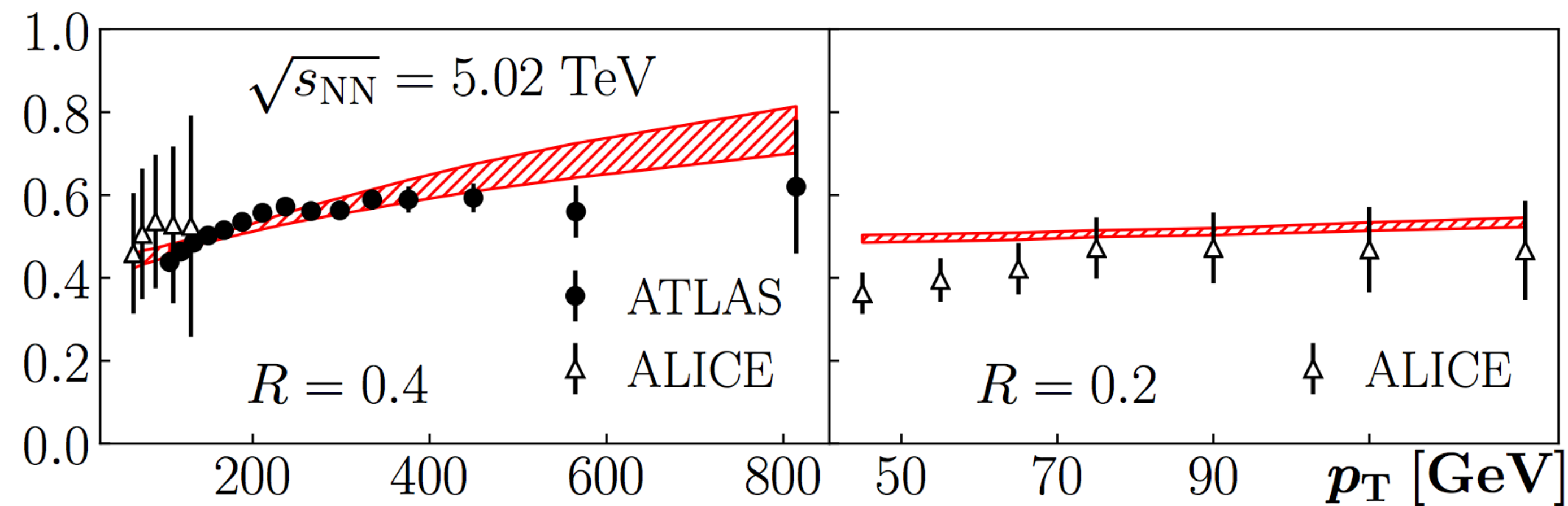
$$\sqrt{s_{NN}} = 2.76 \text{ TeV}$$

$$\chi^2/\text{d.o.f.} = 1.1$$



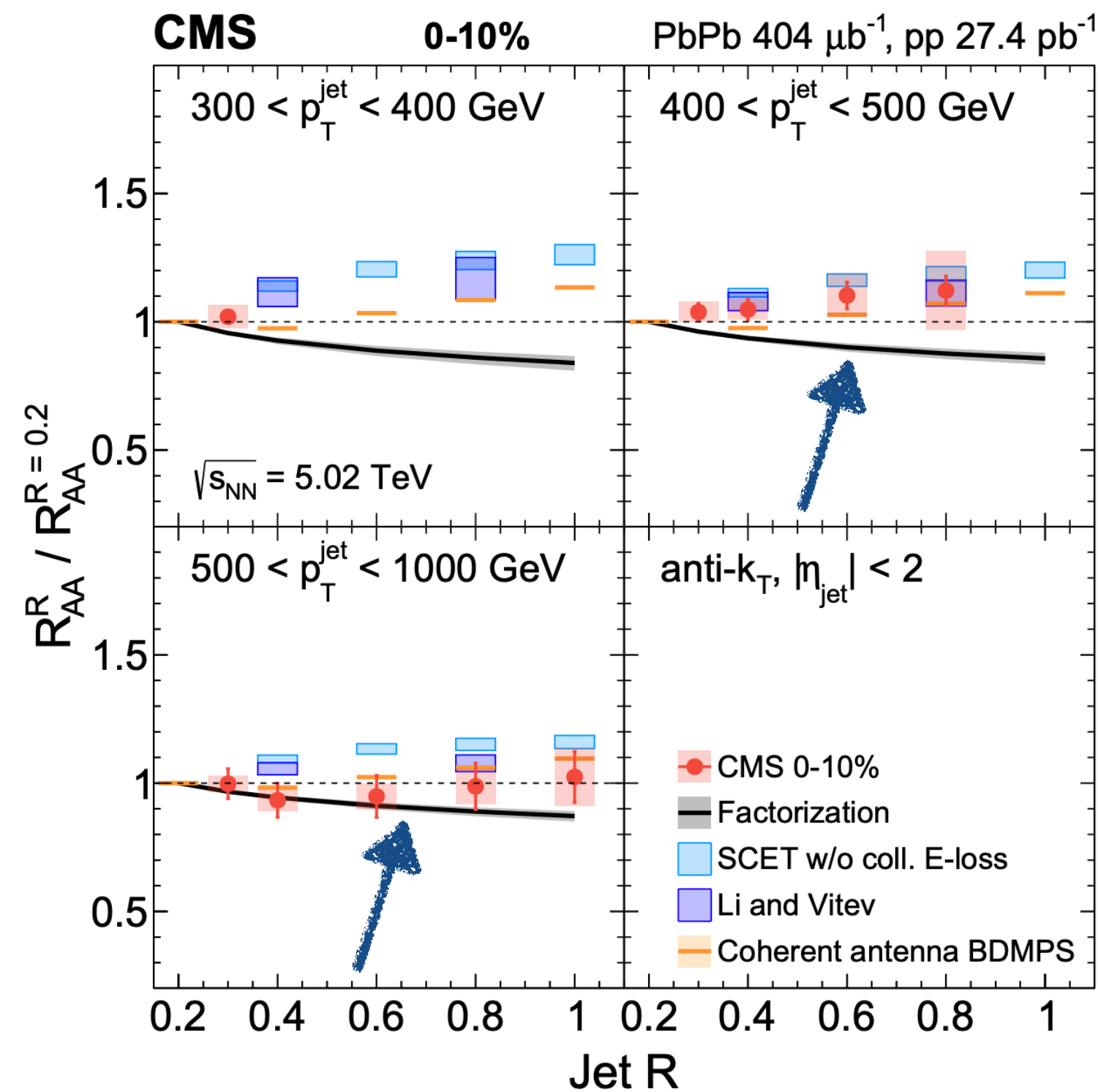
$$\sqrt{s_{NN}} = 5.02 \text{ TeV}$$

$$\chi^2/\text{d.o.f.} = 1.7$$

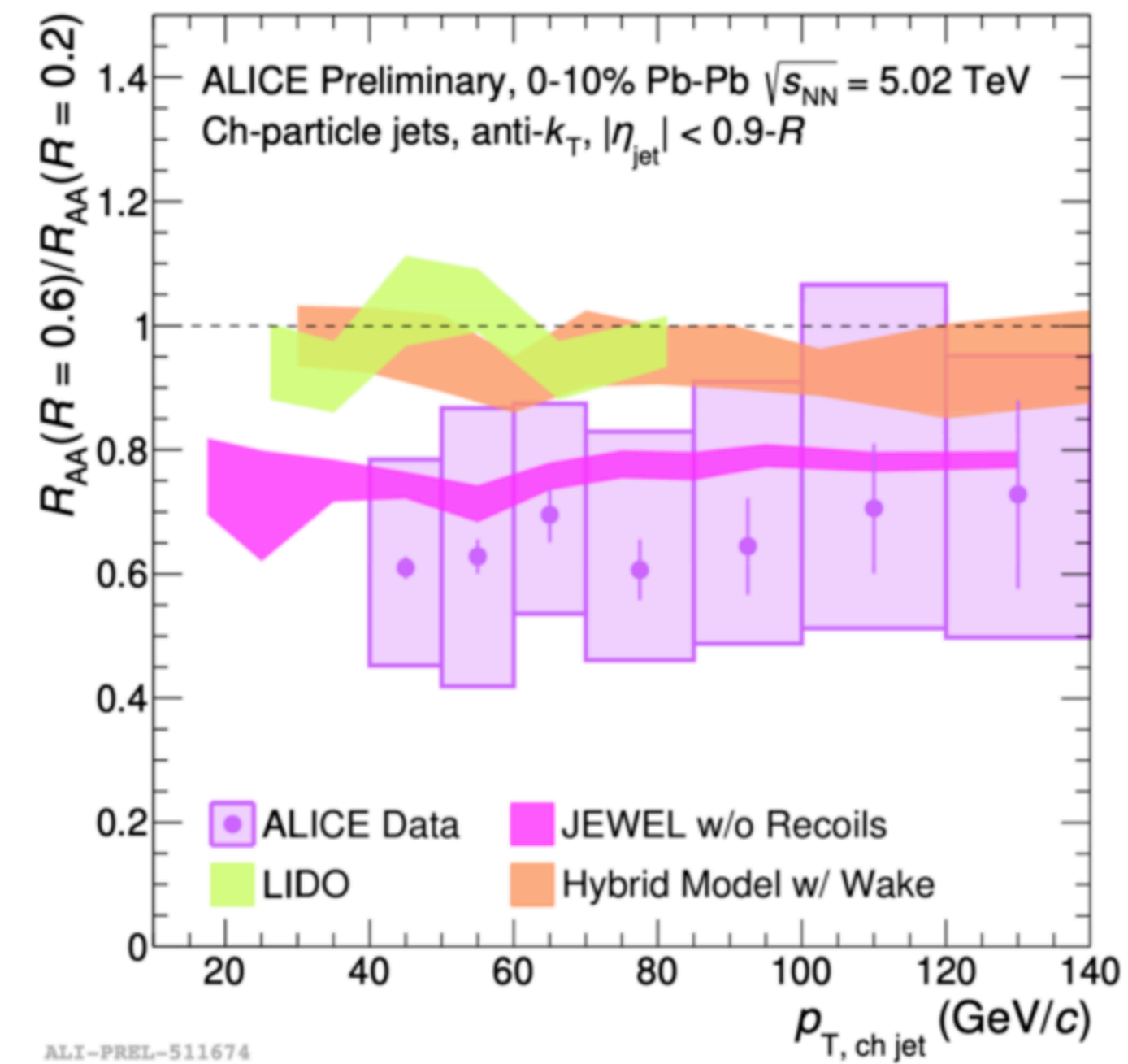


ALICE, PLB 746 (2015) 1  
 ATLAS, PRL 114 (2015) 072302  
 CMS, PRC 96 (2017) 015202  
 ALICE preliminary  
 ATLAS, PLB 790 (2019) 108

# Inclusive jet $R_{AA}$ - radius dependence



CMS, JHEP 05 (2021) 284



<http://alice-figure.web.cern.ch/node/15723/>

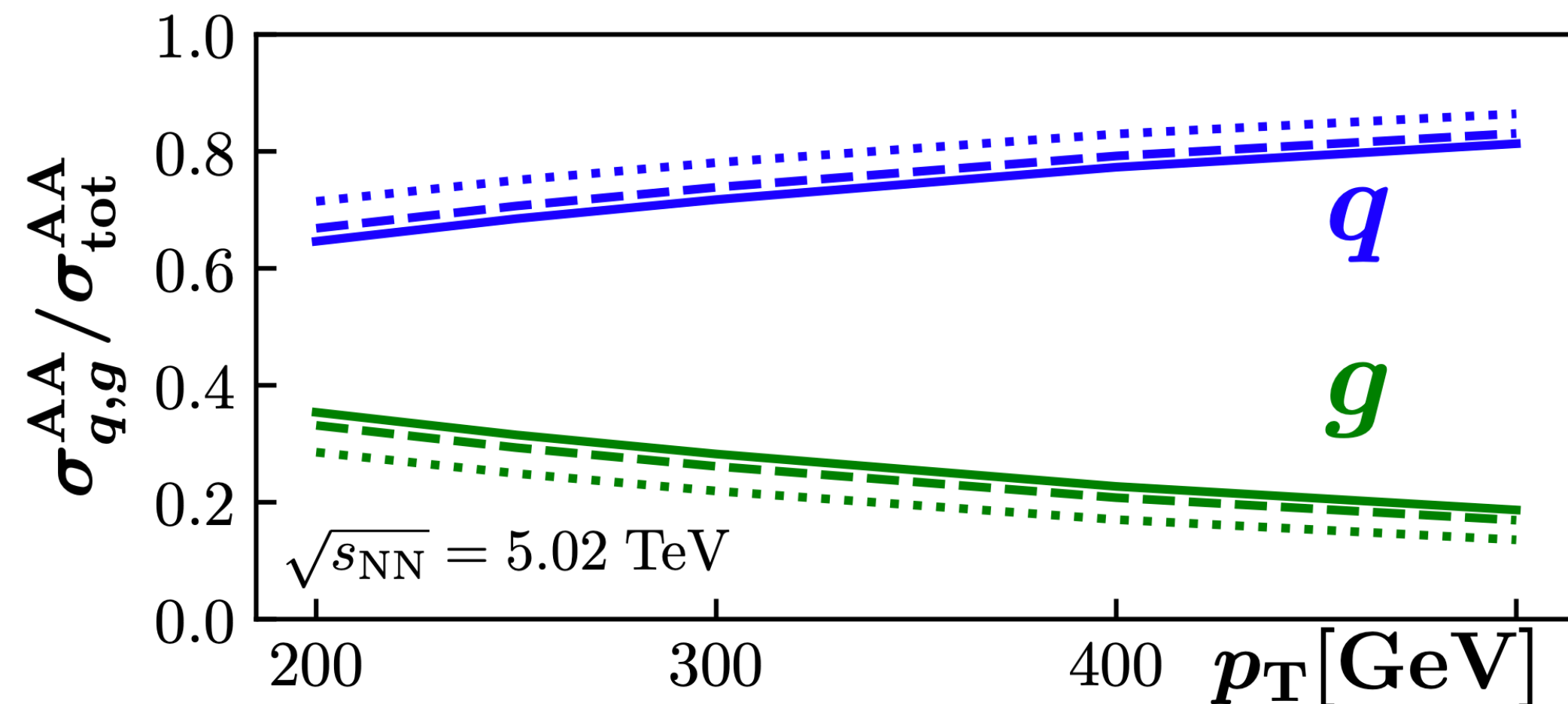
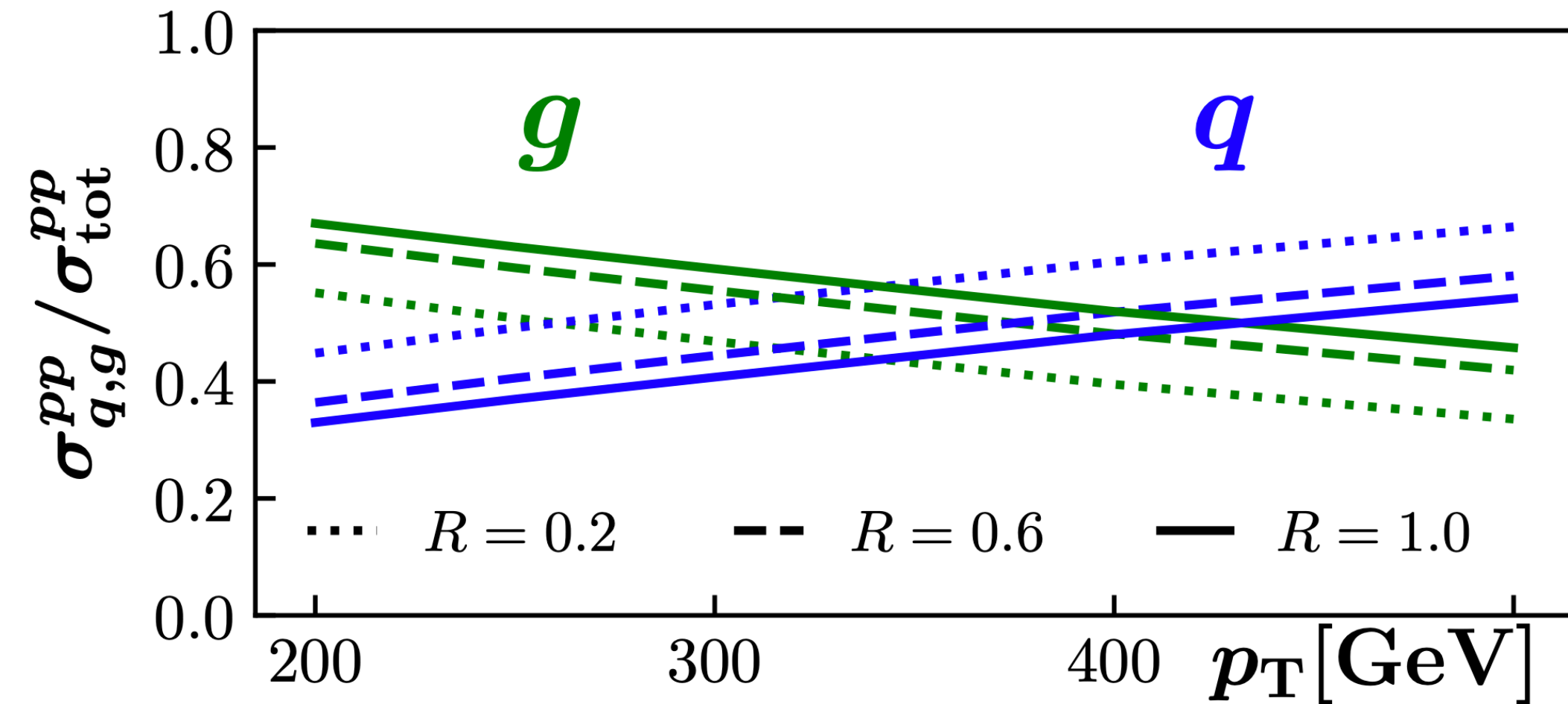
- Data could be included in an updated fit

# In-medium quark/gluon fractions

- Quark/gluon fractions defined at leading power in the jet radius
- Significant shift toward quark jets in the medium

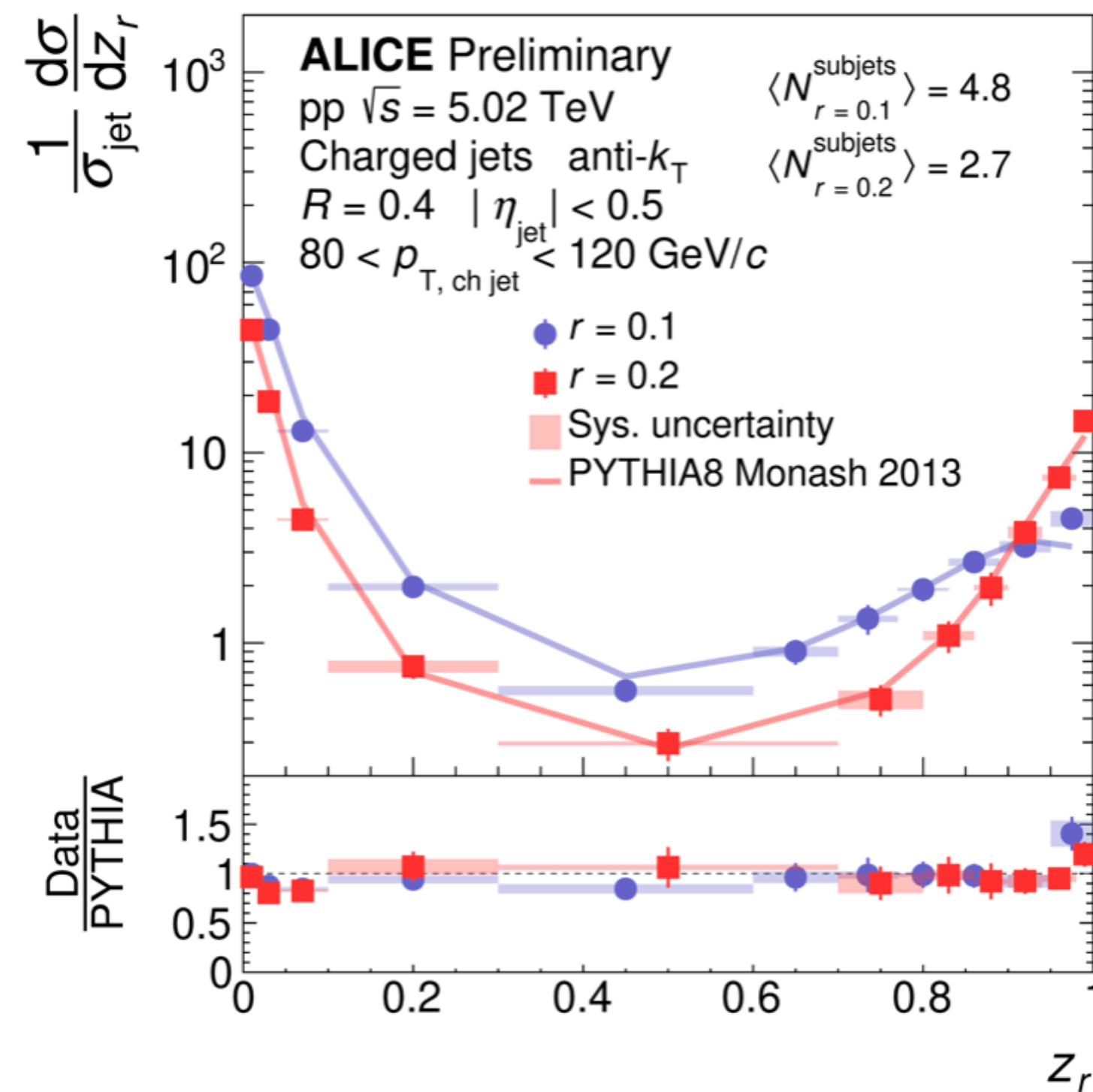


$$f_q J_q(\tau) + f_g J_g(\tau) \quad ?$$



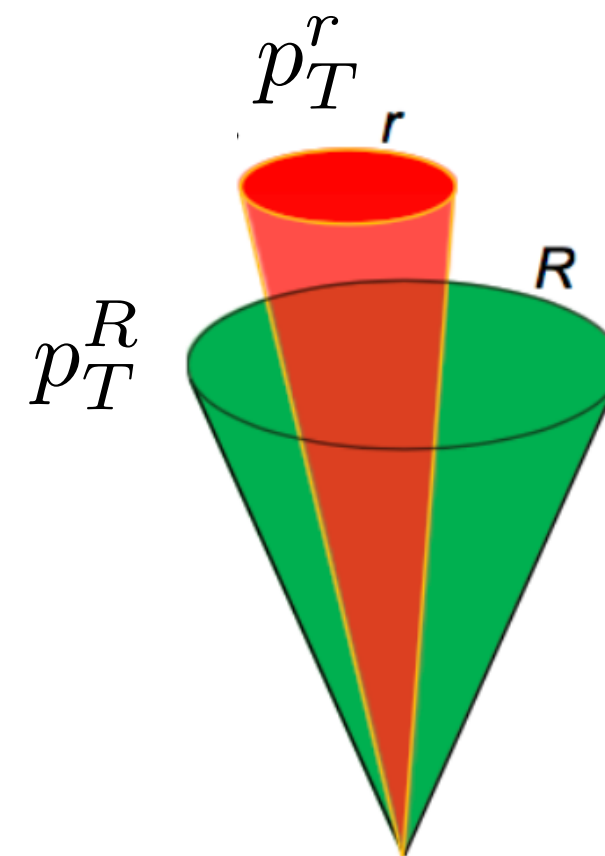
# Applications to jet substructure

- Leading & inclusive subjects in pp



ALI-PREL-490650

$$z_r = p_T^r / p_T^R$$



- Similar factorization in heavy-ion?

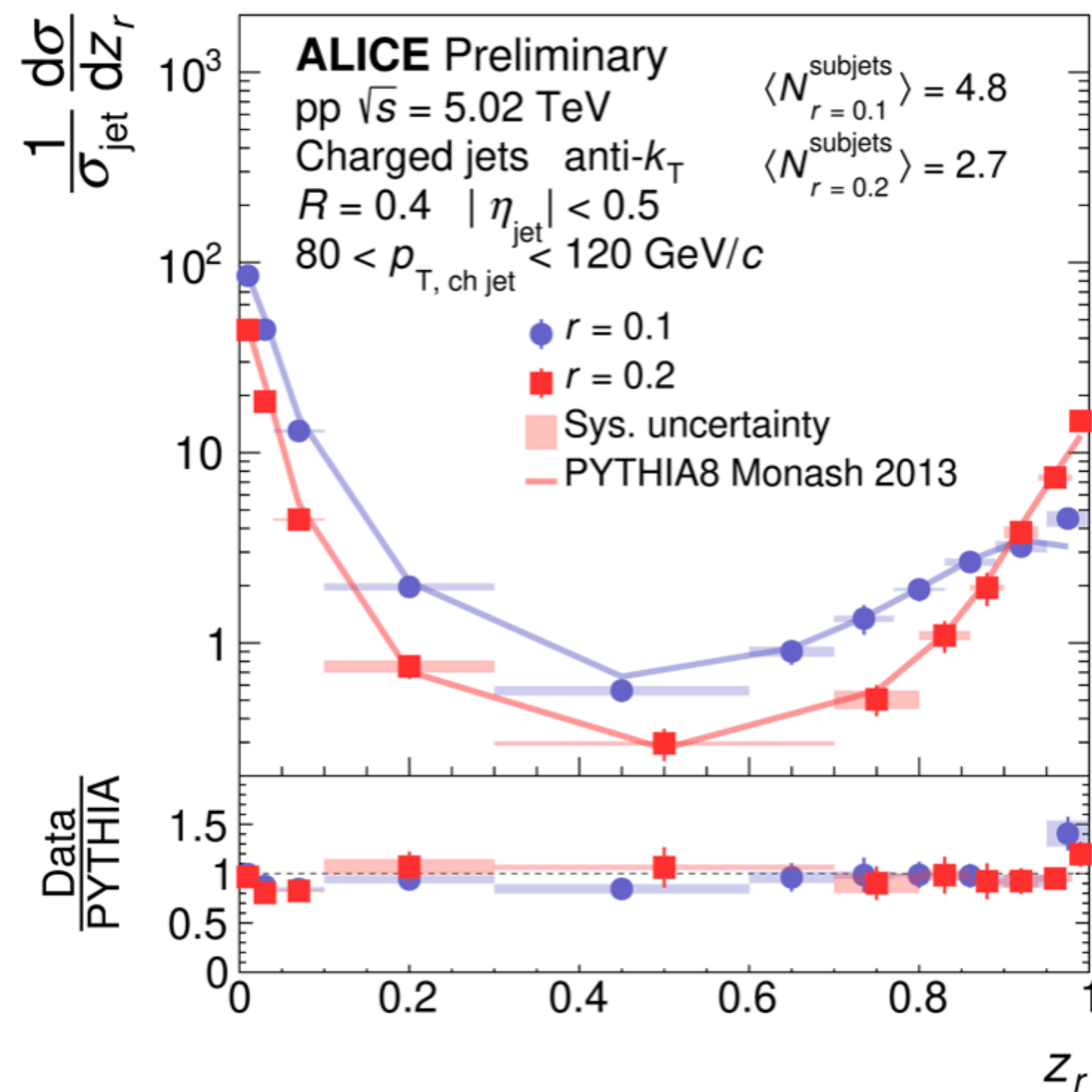
$$\sum_{abcd} f_a \otimes f_b \otimes H_{abc} \otimes J_{cd}^{\text{med}} \times J_d^{\text{med}}(z_r)$$

<https://alice-figure.web.cern.ch/node/19990>

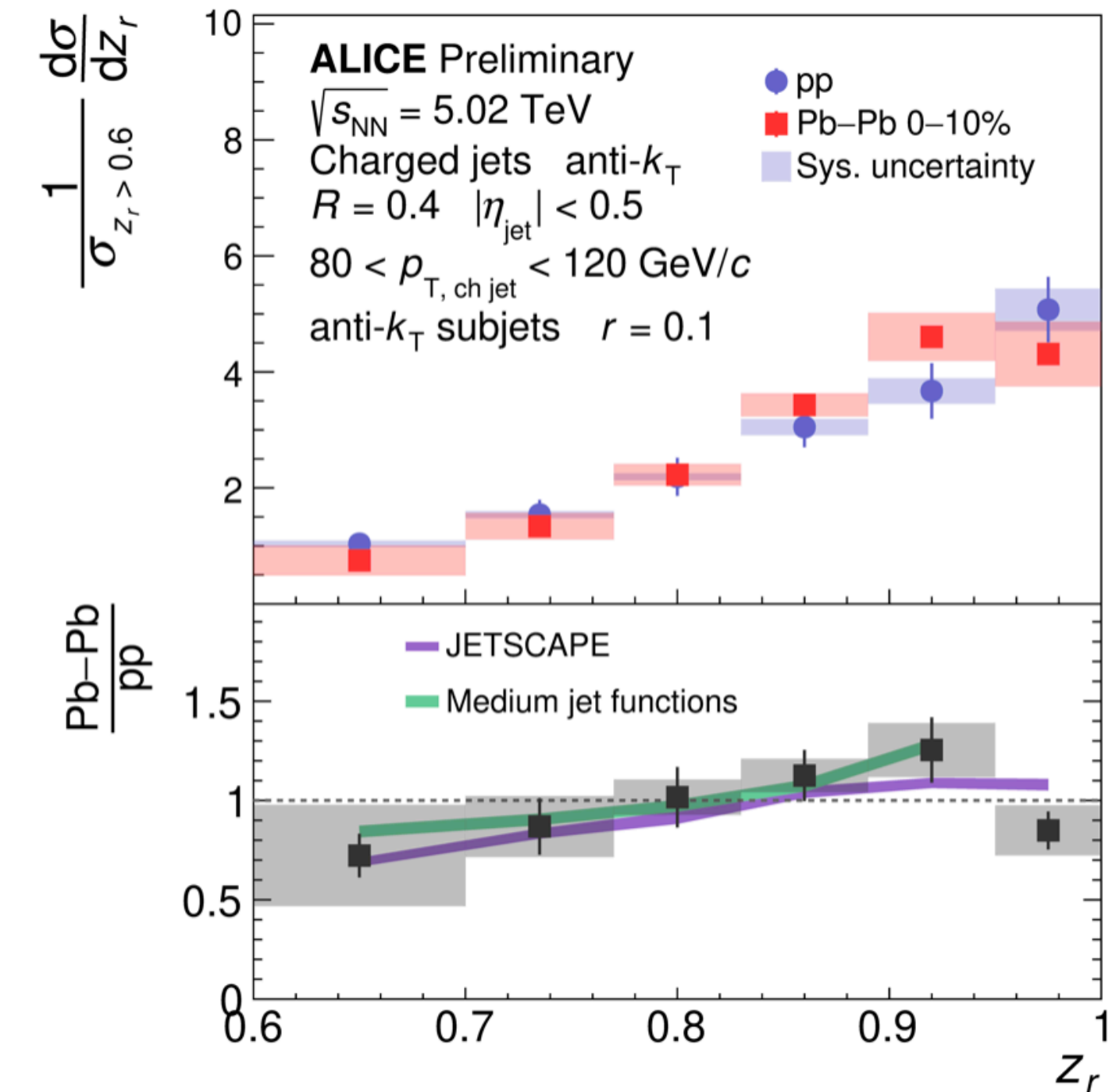
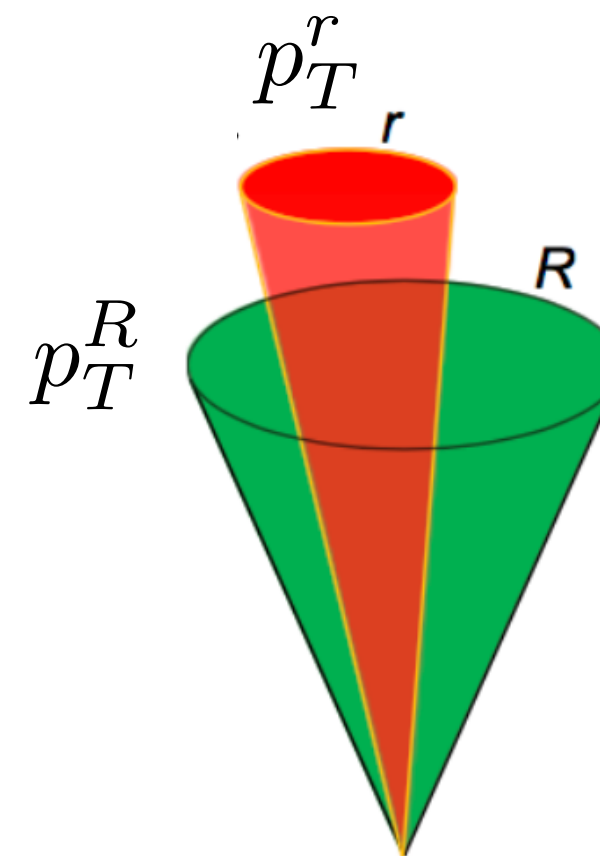
see Dai, Kim, Leibovich `16  
Kang, FR, Waalewijn `17  
Neill, FR, Sato `21

# Applications to jet substructure

- Leading & inclusive subjects in pp



$$z_r = p_T^r / p_T^R$$



ALI-PREL-490655

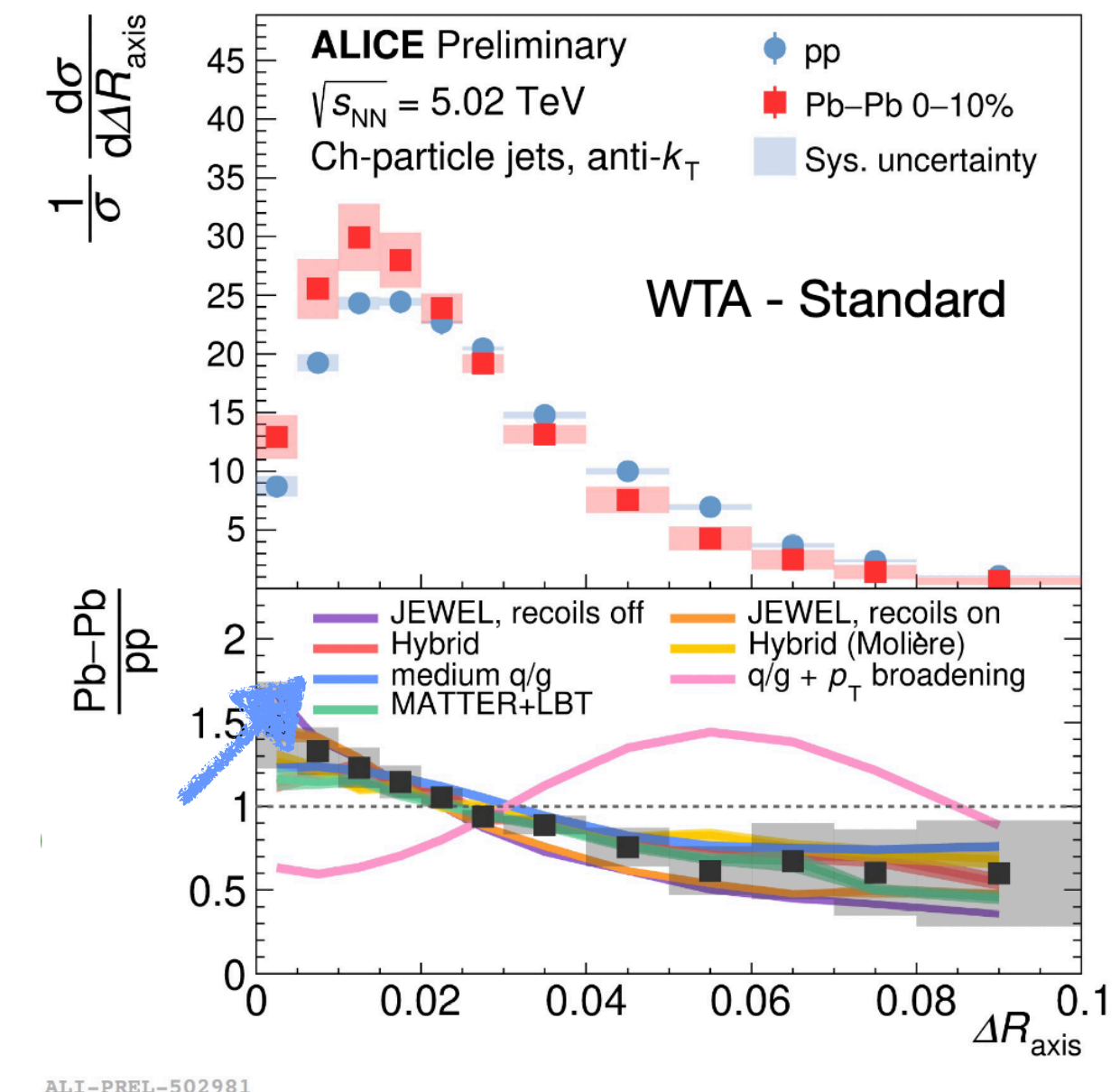
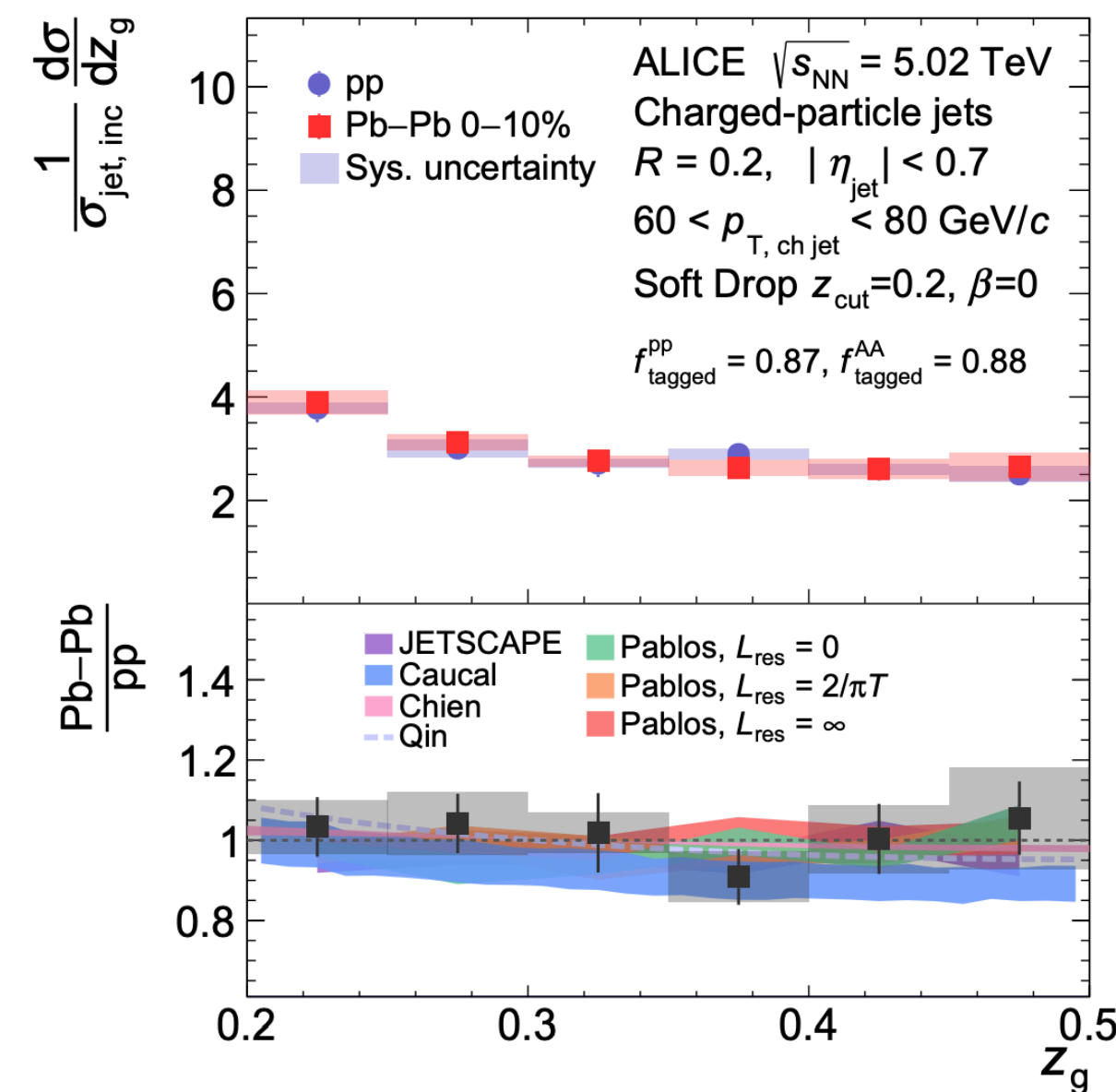
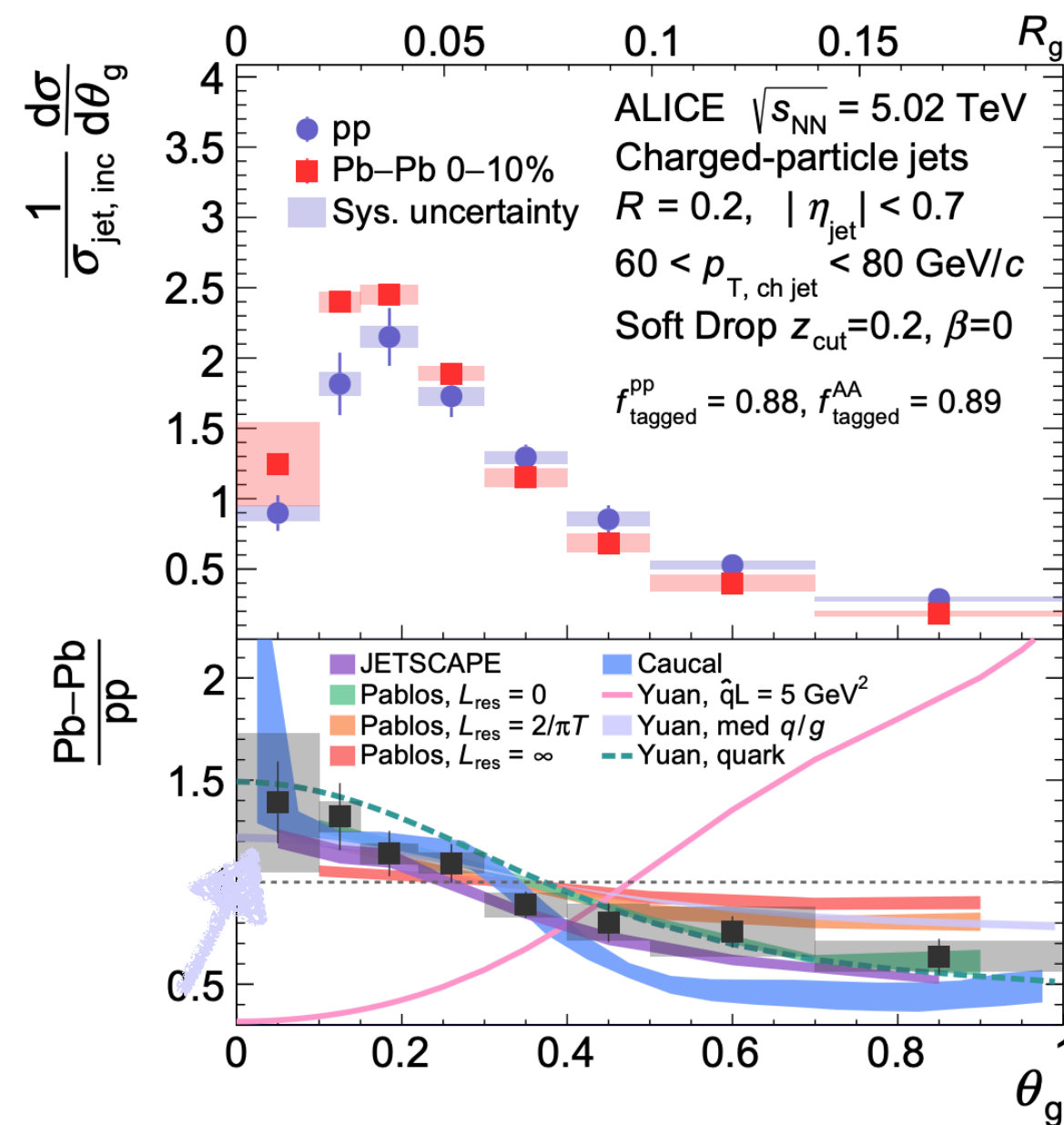
see Dai, Kim, Leibovich `16  
 Kang, FR, Waalewijn `17  
 Neill, FR, Sato `21

<https://alice-figure.web.cern.ch/node/19990>

# Applications to jet substructure

- Soft drop groomed observables

- Angles between jet axes



ALICE, 2107.12984

see also recent ATLAS measurement  
at higher jet  $p_T$

$$f_q J_q(\tau) + f_g J_g(\tau)$$

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# Upper bounds on the information content?

Lai, Mulligan, Ploskon, FR '21

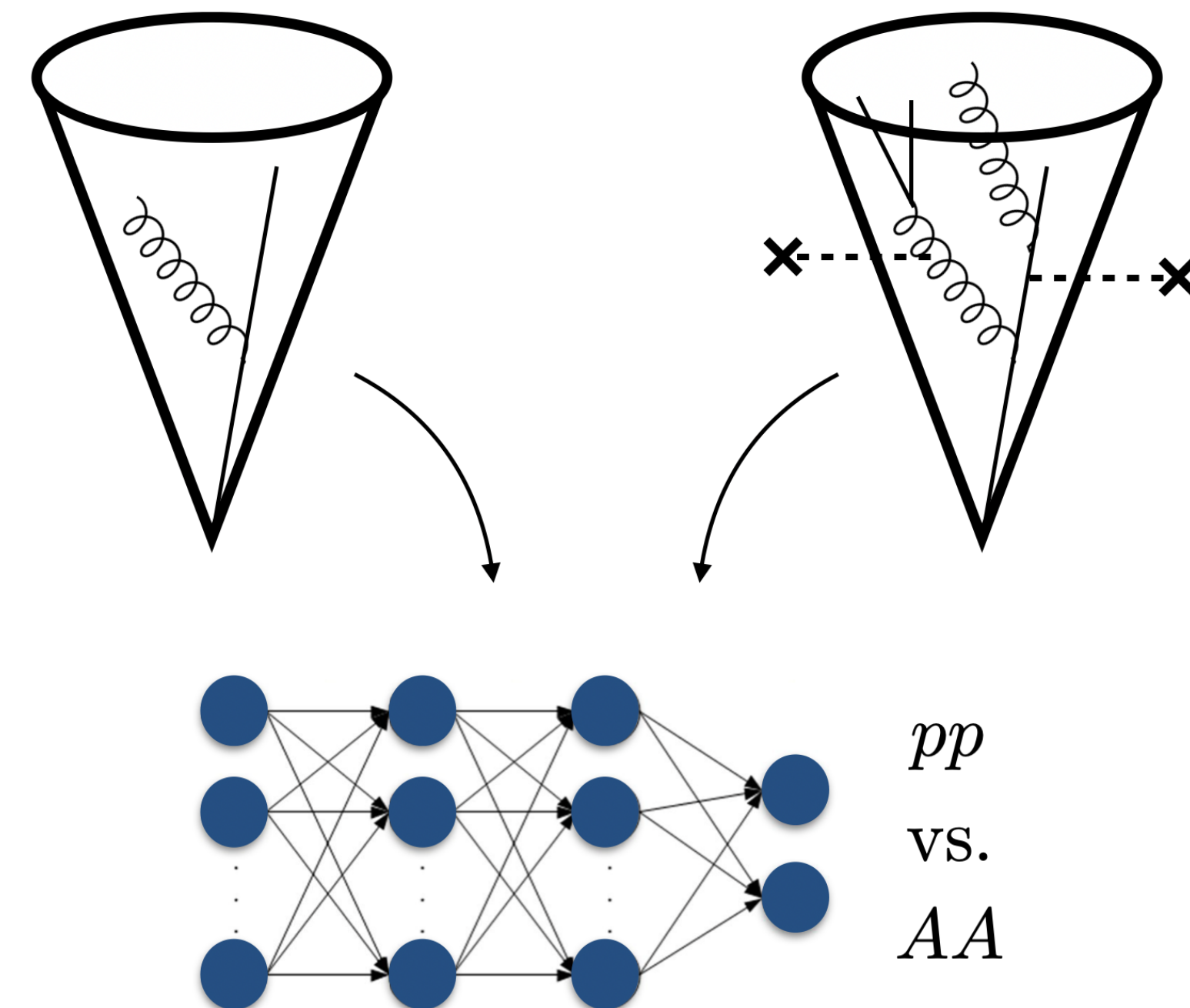
Use machine learning to find an event-by-event “observable”



Can formulate as a binary classification problem



Compare performance of classifier to traditional observables

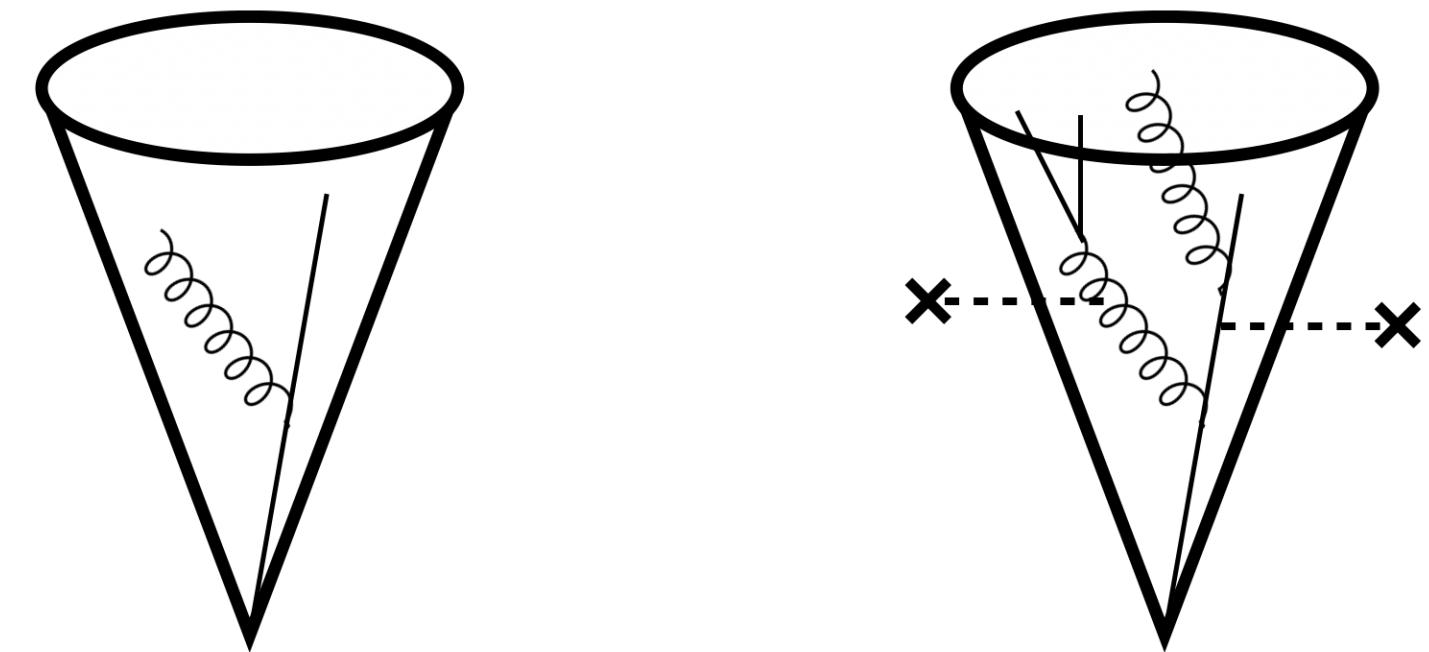




# Upper bounds on the information content?

Lai, Mulligan, Ploskon, FR '21

- pp & AA jets generated in Pythia/Jewel
- Use deep sets that are known to be one of the best quark/gluon classifiers



$$f(p_1, \dots, p_M) = F \left( \sum_{i=1}^M \Phi(p_i) \right)$$

Classifier

DNNs

Zaheer et al. 1703.06114  
Wagstaff et al. 1901.09006  
Bloem-Reddy, Teh JMLR 21 90 (2020)

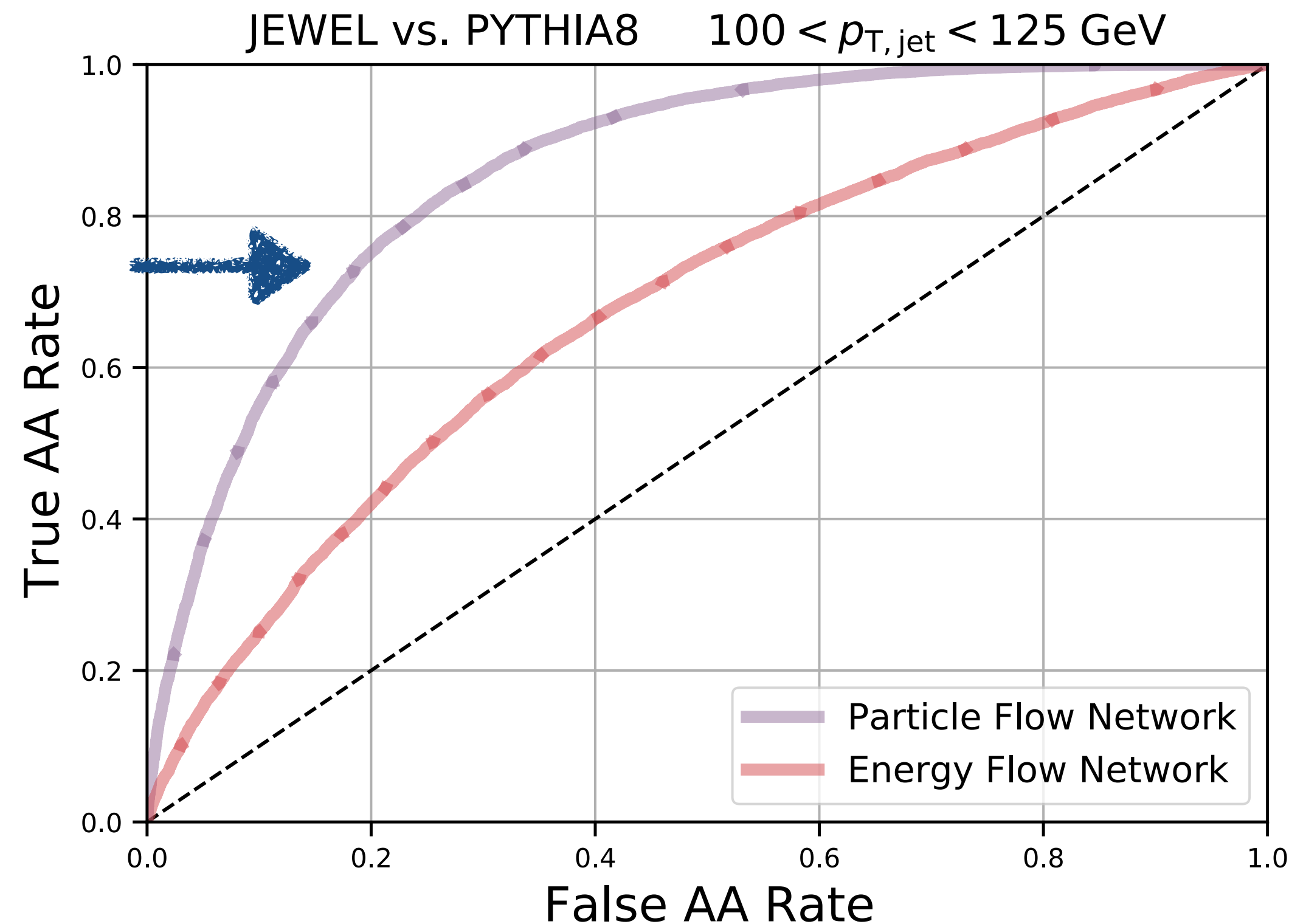
Input is the set of particle 4-momenta

Komiske, Metodiev, Thaler '19

# Upper bounds on the information content?

Lai, Mulligan, Ploskon, FR '21

- Quantify performance using ROC curves



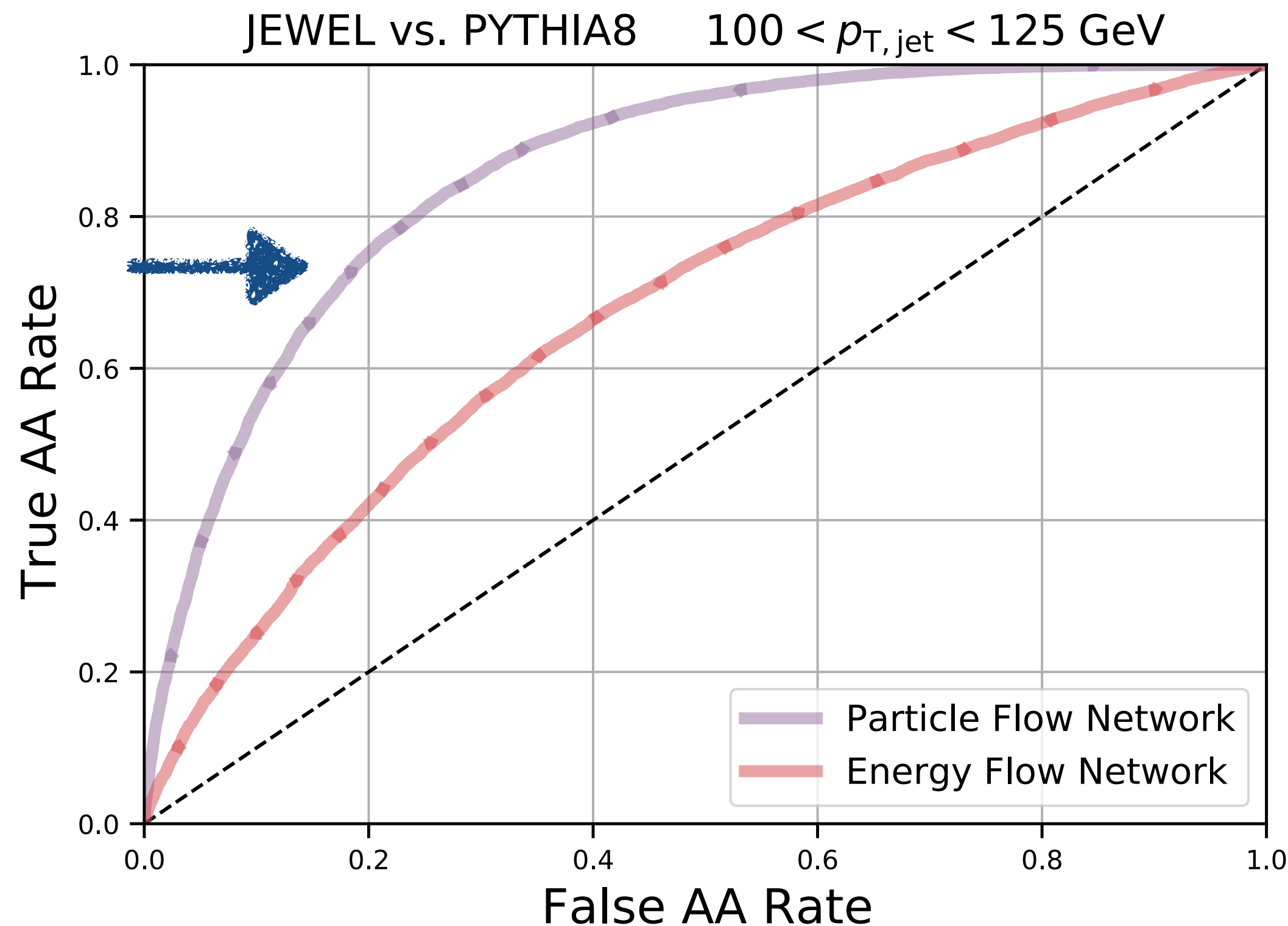
- ML significantly outperforms traditional observables
- There is a lot more information in quenched jets that we are currently not making use of

See also Chien, Elayavalli; Apolinario, Castro, Romao, Milhano, Pedro, Peres; Liu, Velkovska, Verweij

# Upper bounds on the information content?

Lai, Mulligan, Ploskon, FR '21

- Quantify performance using ROC curves



- ML significantly outperforms traditional observables
- There is a lot more information in quenched jets that we are currently not making use of
- Can we identify observables to close the information gap?

# How to systematically use all the information?

Lai, Mulligan, Ploskon, FR '21

- Can use complete sets of observables

- Basis of N-subjettiness observables *Datta, Larkoski '17 - '19*

$$\tau_N^{(\beta)} = \frac{1}{p_T^{\text{jet}}} \sum_{i \in \text{Jet}} p_{Ti} \min \left\{ R_{1i}^\beta, R_{2i}^\beta, \dots, R_{Ni}^\beta \right\}$$

2-body :  $\tau_1^{(1)}, \tau_1^{(2)}$ ,

3-body :  $\tau_1^{(0.5)}, \tau_1^{(1)}, \tau_1^{(2)}, \tau_2^{(1)}, \tau_2^{(2)}$ ,

⋮

M-body :  $\tau_1^{(0.5)}, \tau_1^{(1)}, \tau_1^{(2)}, \dots, \tau_{M-2}^{(0.5)}, \tau_{M-2}^{(1)}, \tau_{M-2}^{(2)}, \tau_{M-1}^{(1)}, \tau_{M-1}^{(2)}$ .

- Basis of Energy Flow Polynomials

*Komiske, Metodiev, Thaler '17,  
Cal, Thaler, Waalewijn '22*

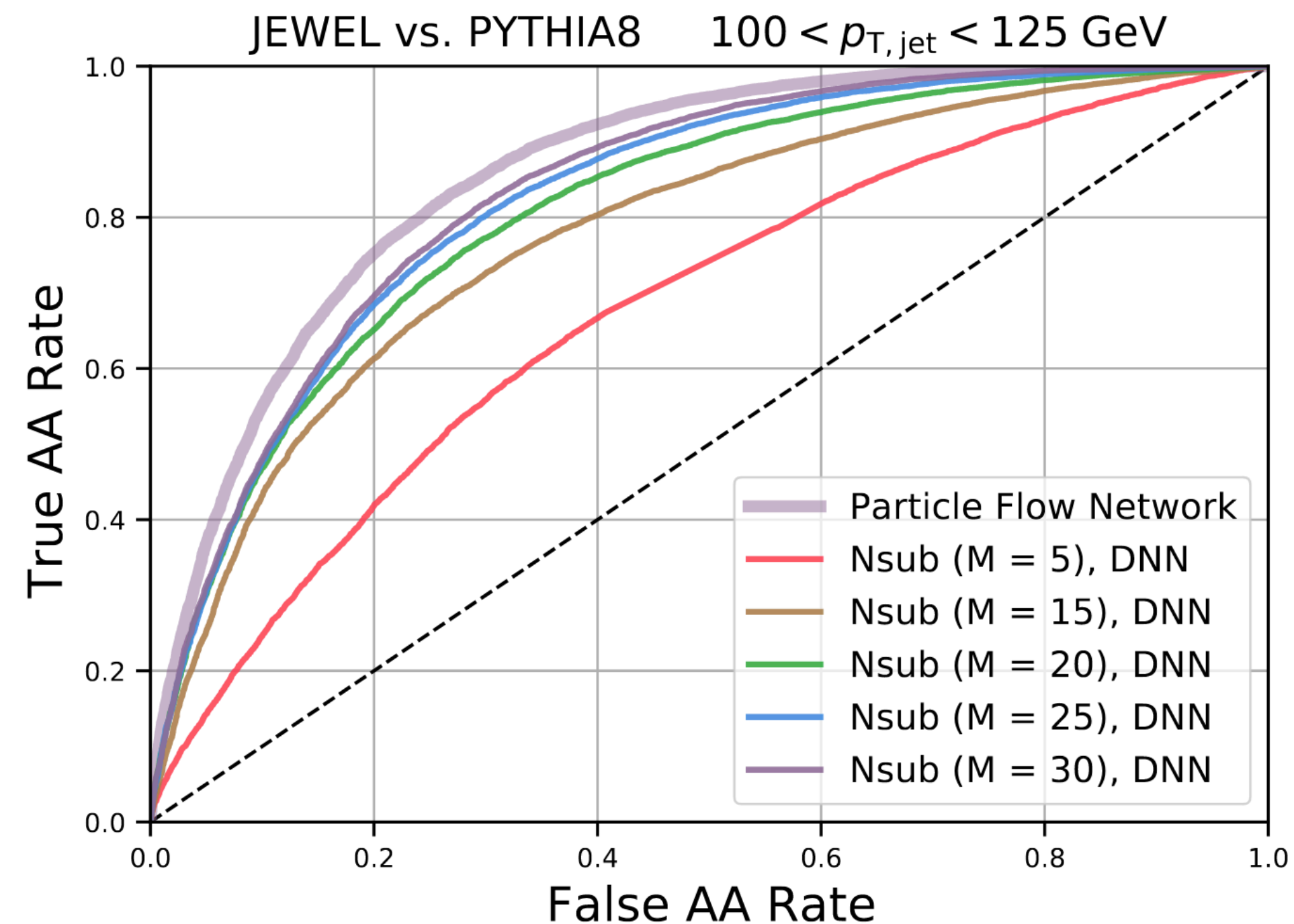
$$\text{EFP}_G = \sum_{i_1=1}^M \cdots \sum_{i_V=1}^M z_{i_1} \cdots z_{i_V} \prod_{(k,l) \in E} \theta_{i_k i_l}$$



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Lai, Mulligan, Ploskon, FR '21

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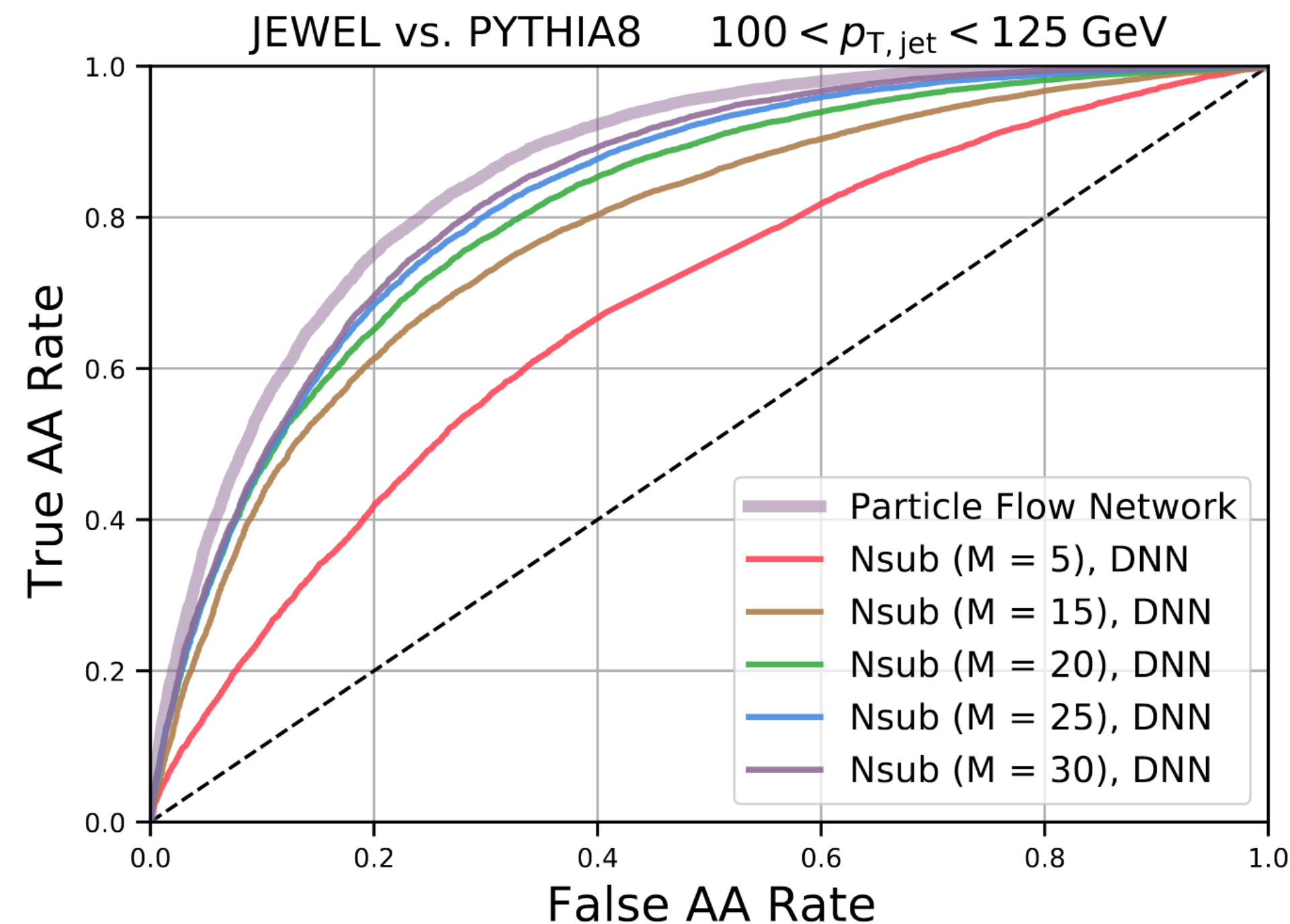


- Allows us to systematically approximate the full information content

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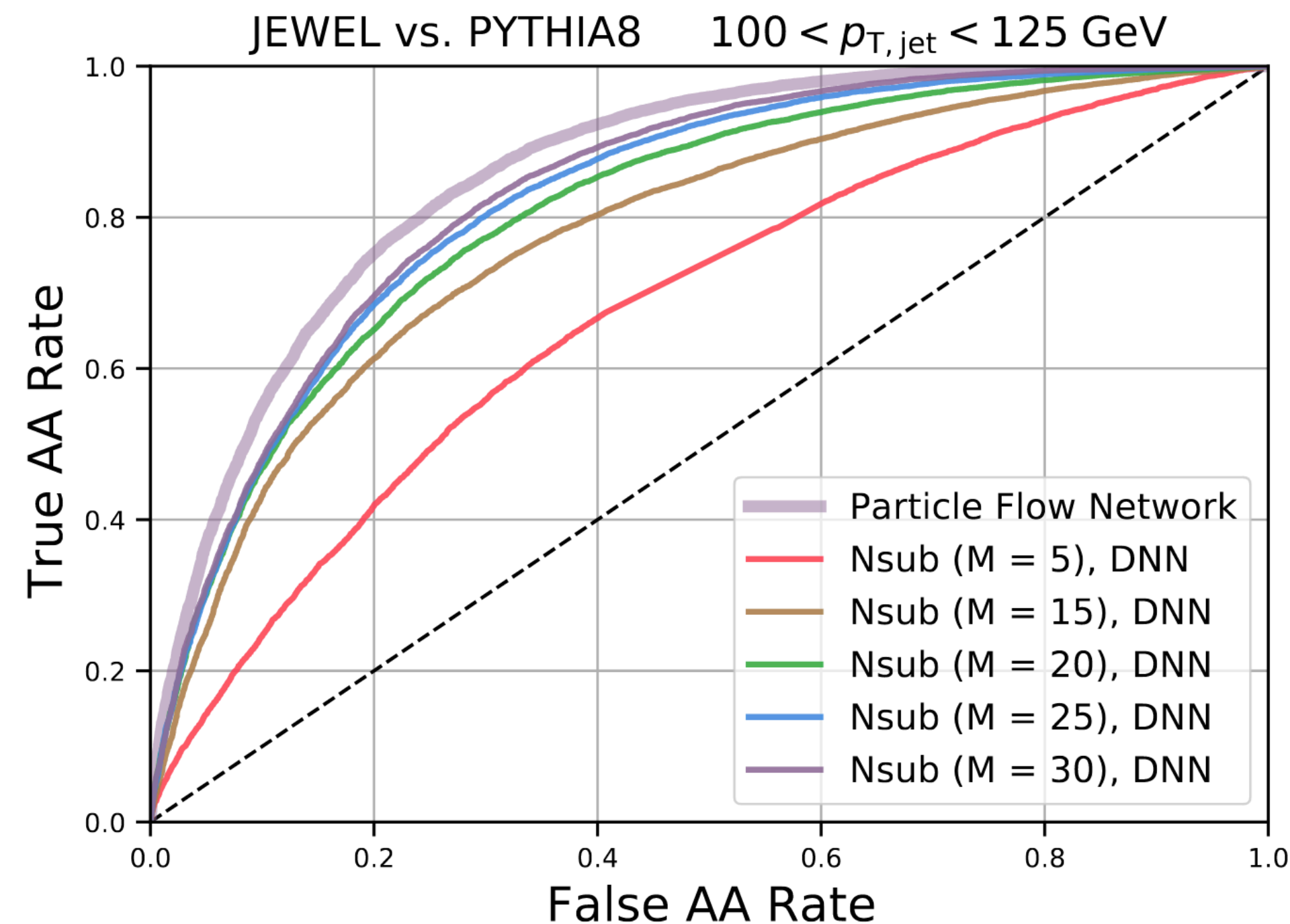


- Allows us to systematically approximate the full information content
- Using symbolic/Lasso regression we can identify the most important ones

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- Allows us to systematically approximate the full information content
- Using symbolic/Lasso regression we can identify the most important ones
- Heavy-ion background & analysis needs to be performed on data instead of simulations

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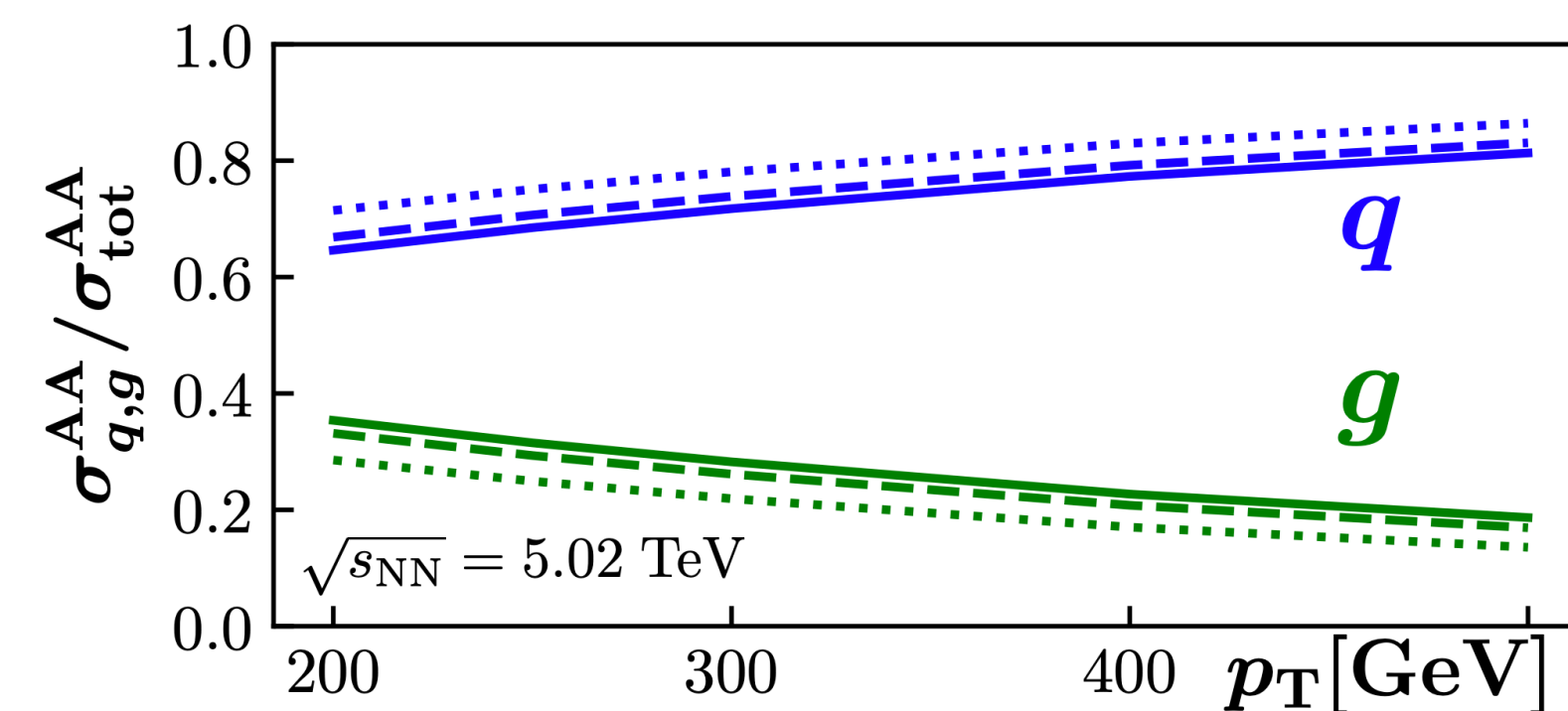
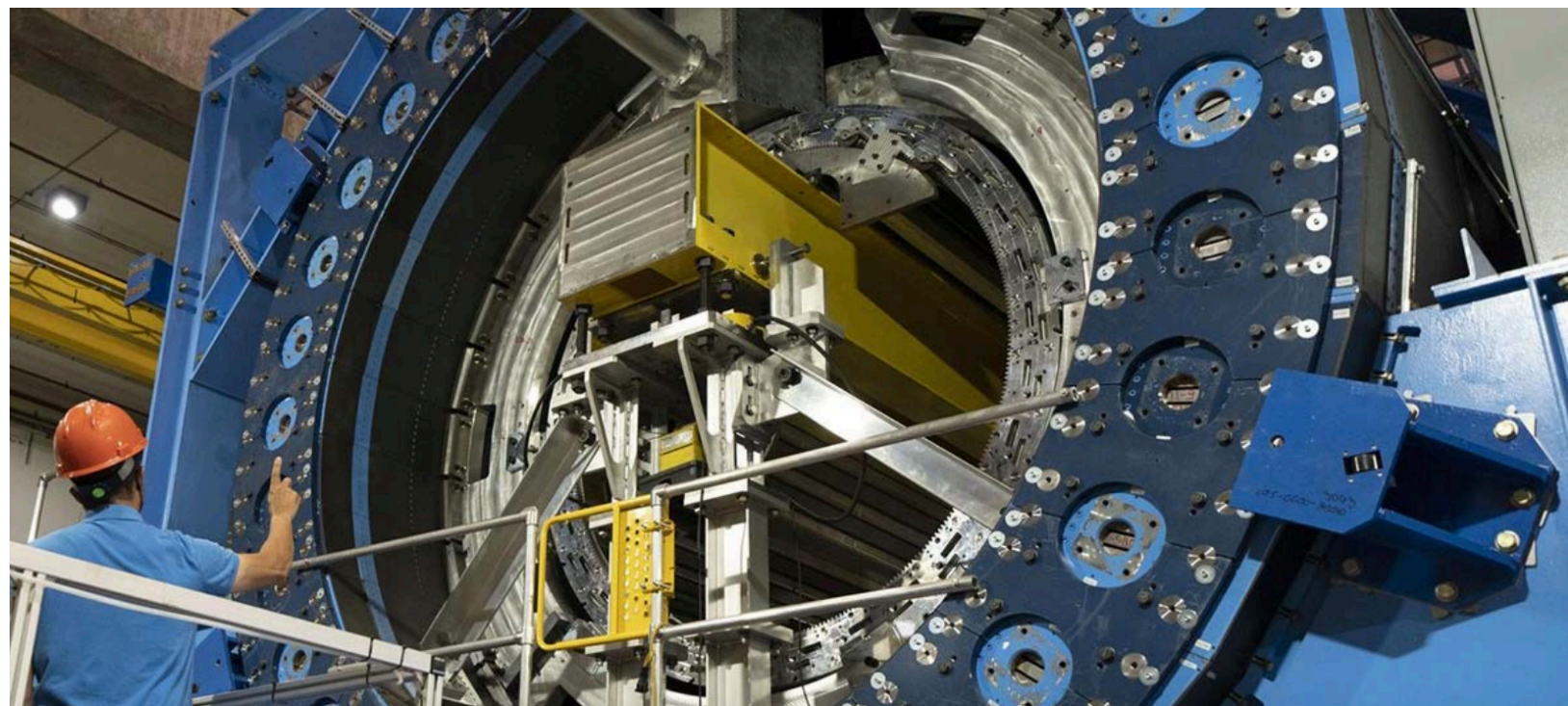
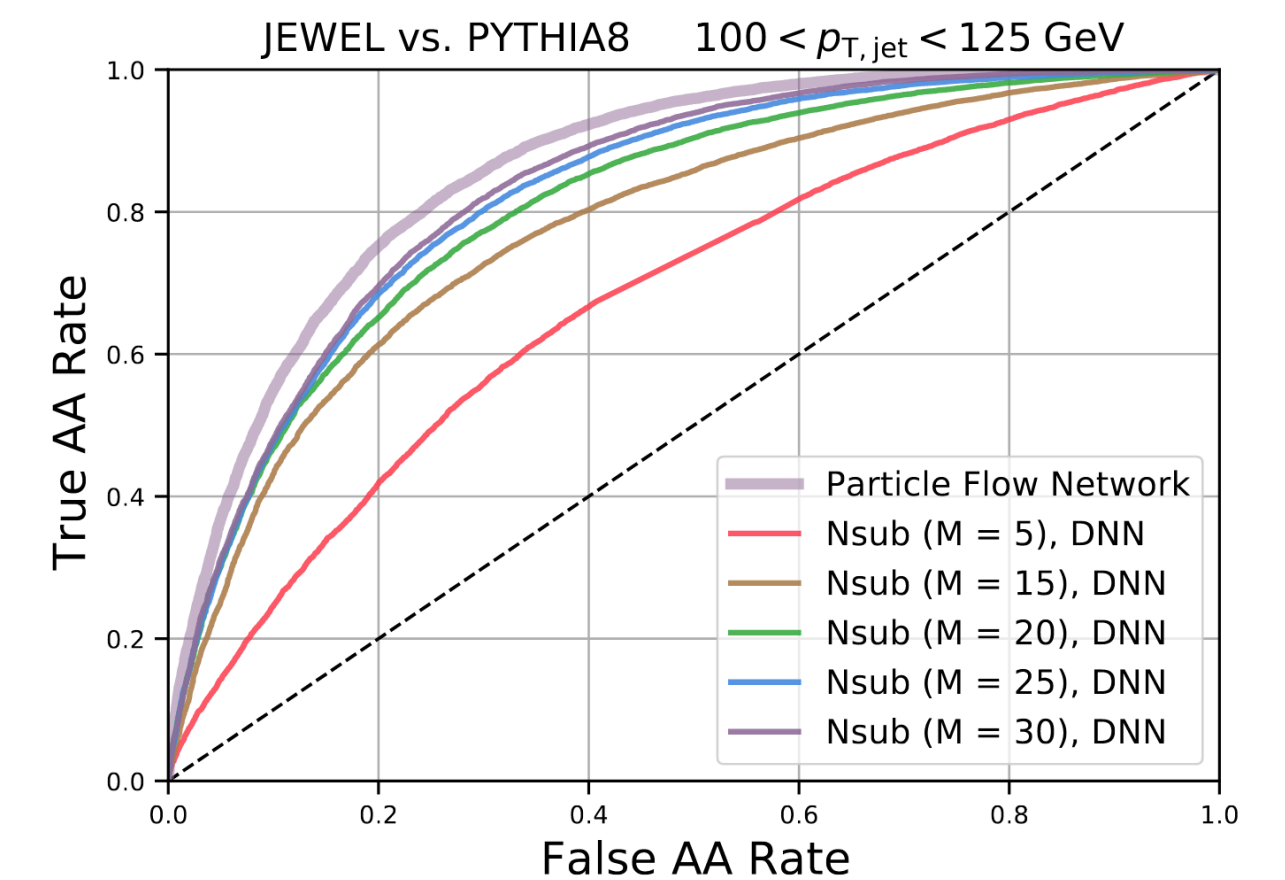
Inclusive jets &  
quark/gluon fractions

Machine-learned  
jet substructure  
observables



# Conclusions

- Factorization and universality at RHIC energies with sPHENIX?
- Further theory developments needed
- Information content of jet substructure observables?
- Improved extraction of medium properties?
- Analyze data using ML?

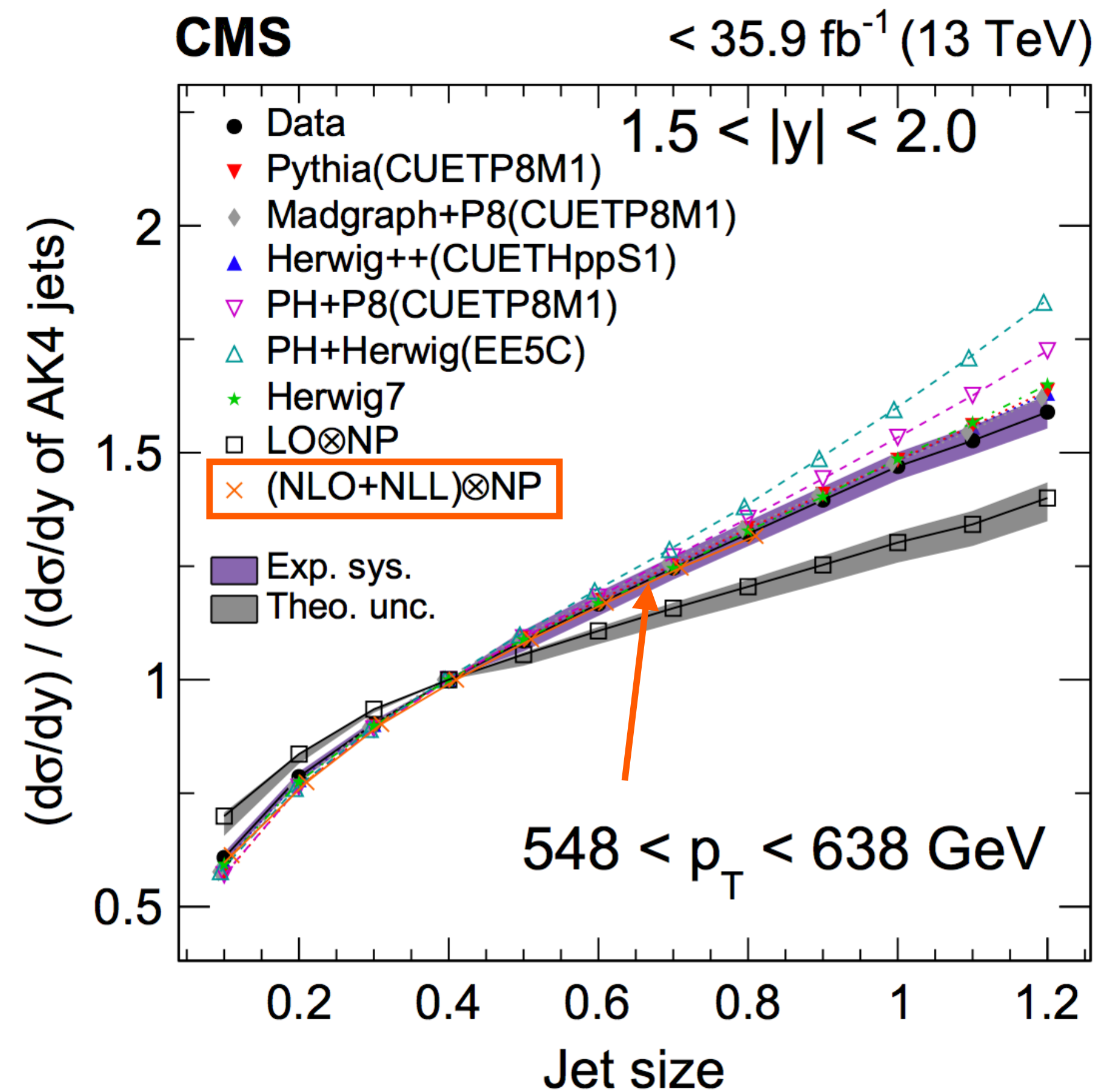


# Inclusive jets in proton-proton collisions

- Jet radius dependence of the inclusive jet cross section

CMS, JHEP 12 (2020) 82  
see also recent results from ALICE

Liu, Moch, FR '17, 18

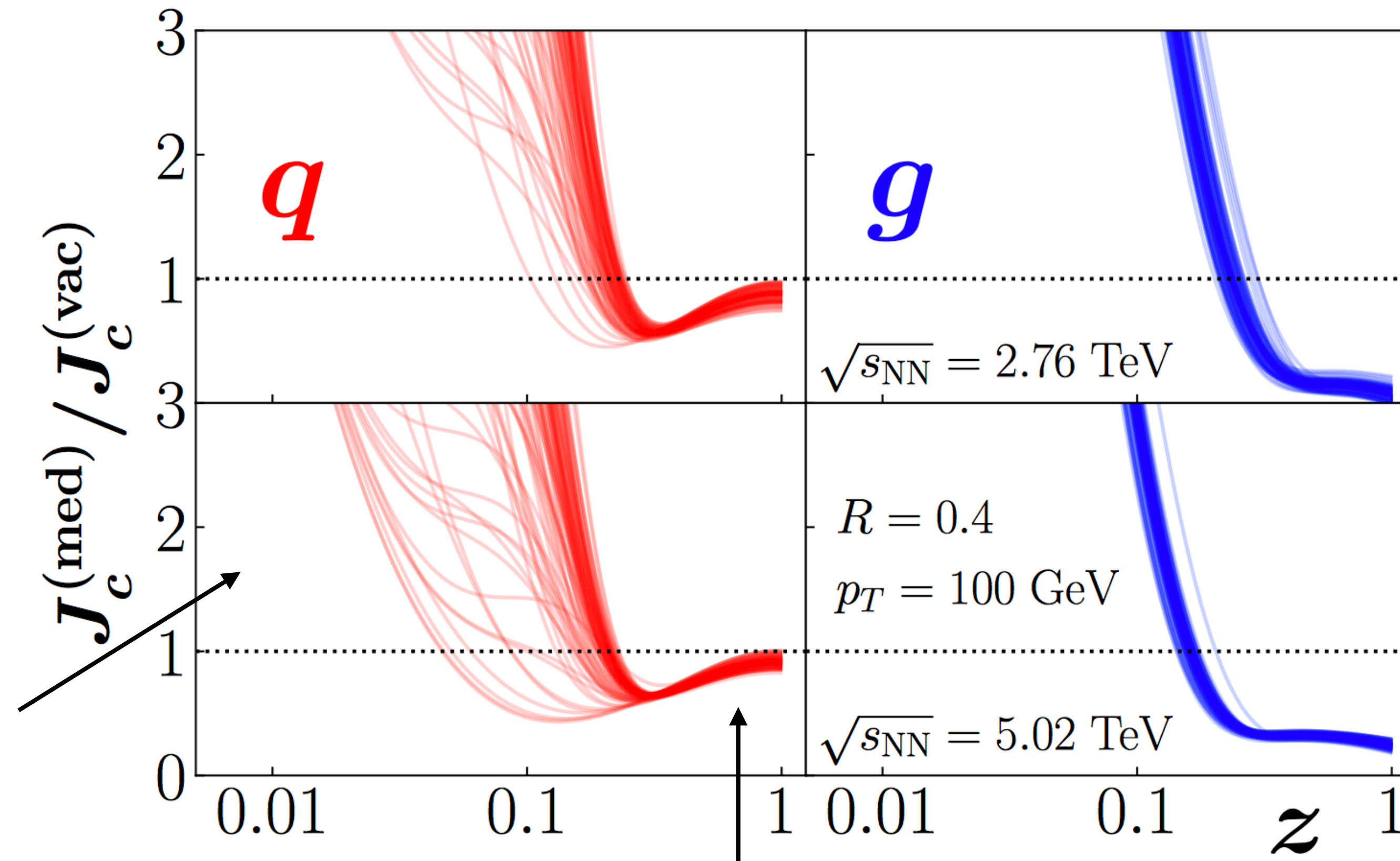


# In-medium jet functions

Qiu, FR, Sato, Zurita '19

- Suppression at large- $z$  compensated for by enhancement at small- $z$

Small- $z$  region generally less constrained



Potentially requires threshold resummation for  $z \rightarrow 1$