

# How to better compare pp and AA data?

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HOLMGANGA

# Introduction

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We are all interested in understanding the LHC observations in pp collisions: collective-like behaviour, charm/bottom baryon-to-meson ratios and similarities with the light flavour sector, strings vs QGP scenarios, ... BUT there are some basic questions that we have not fully addressed, their answers may be important to understand the data

# Introduction

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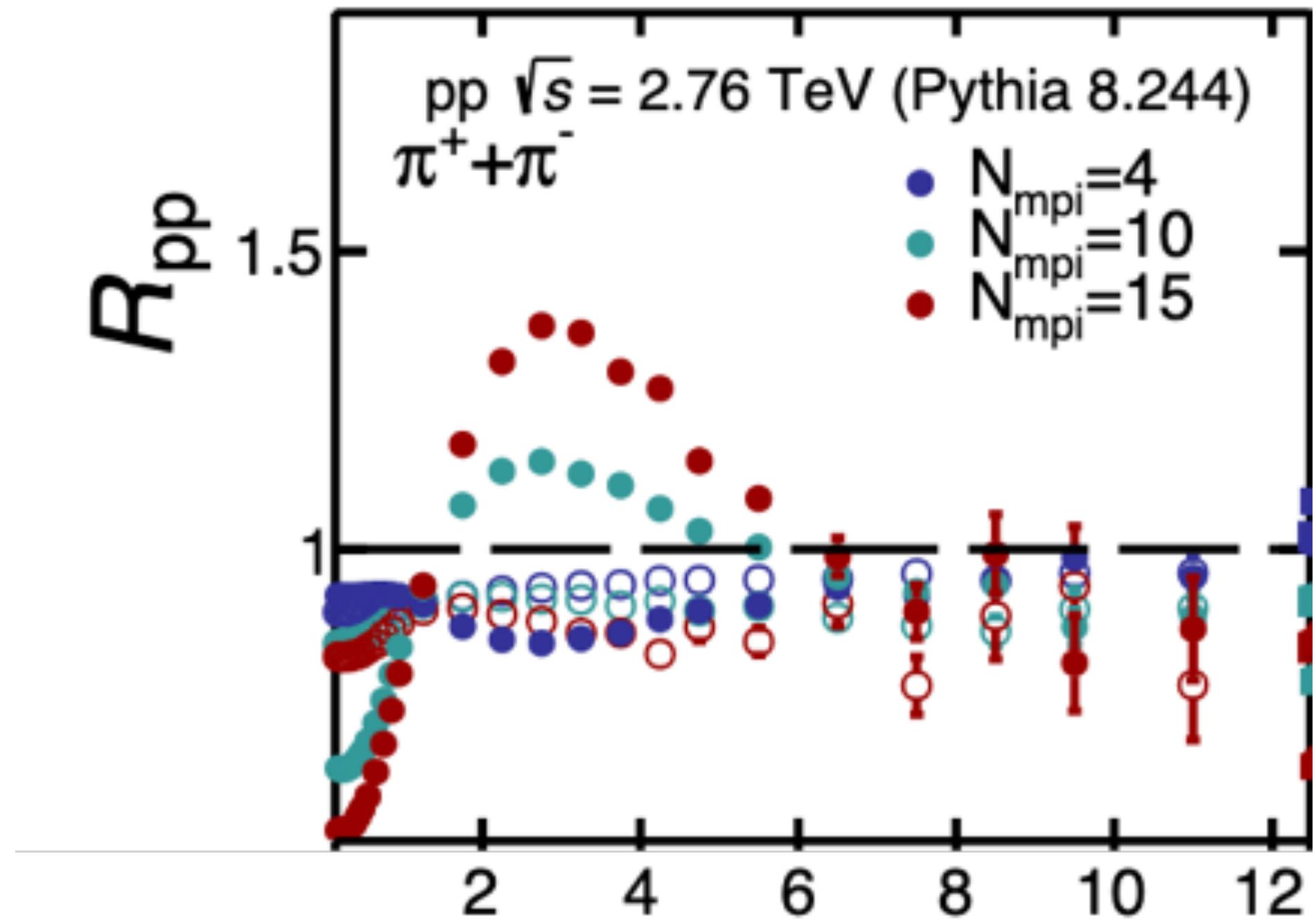
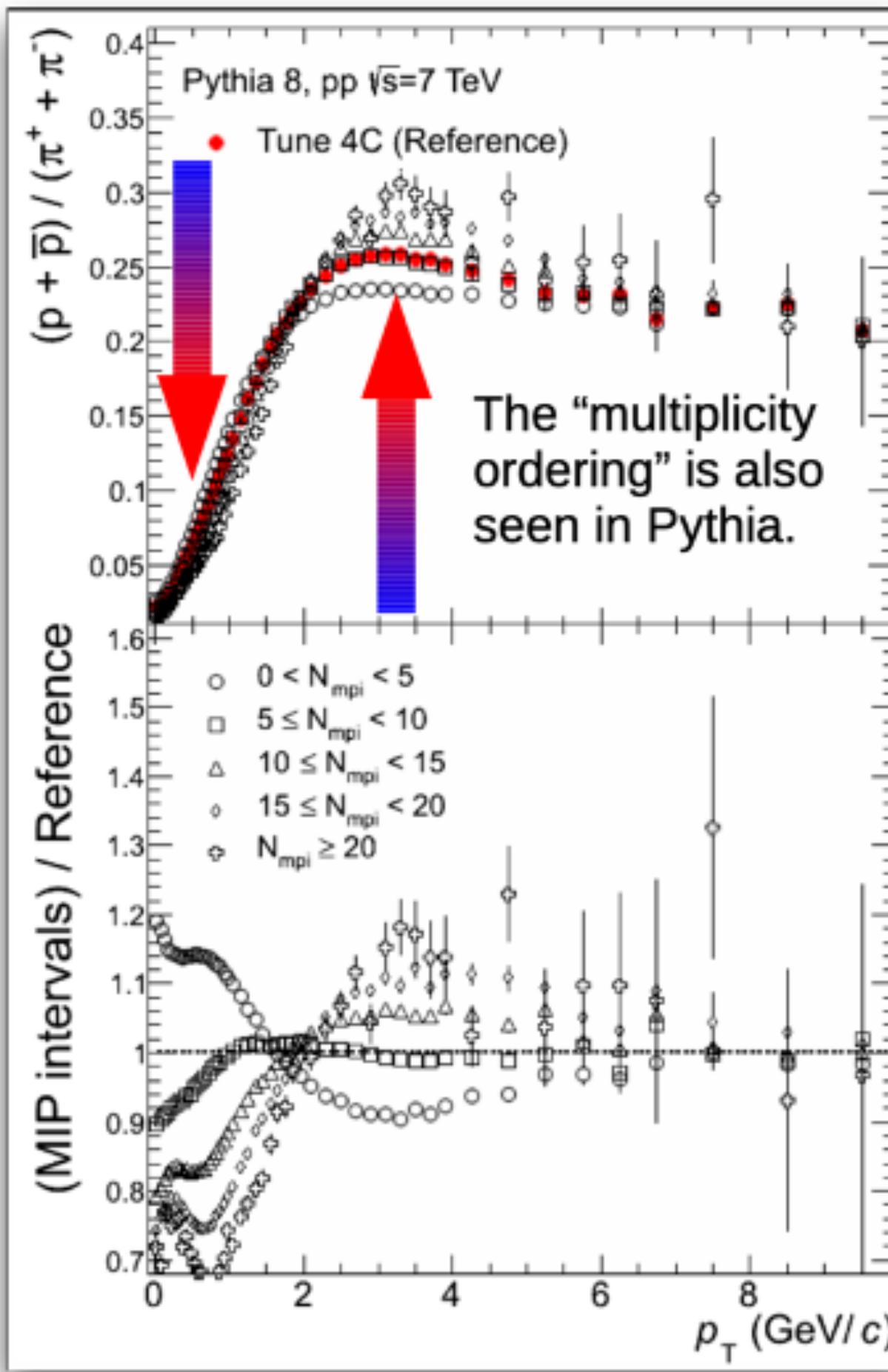
We are all interested in understanding the LHC observations in pp collisions: collective-like behaviour, charm/bottom baryon-to-meson ratios and similarities with the light flavour sector, strings vs QGP scenarios, ... BUT there are some basic questions that we have not fully addressed, their answers may be important to understand the data

- What type of pp collisions should we compare with heavy-ion collisions (HIC)?
- How can we tag those pp collisions (event classifiers)
- Do we understand the bias of the different classifiers explored so far?
- How are they correlated?
- Are our data biased in the same way as MC?

# What type of pp collisions should we compare with HIC?

- Collisions with several “soft” parton-parton scatterings seems like the most natural candidates. Example using Pythia (MPI + CR)

[Phys. Rev. Lett. 111, 042001 \(2013\)](#)

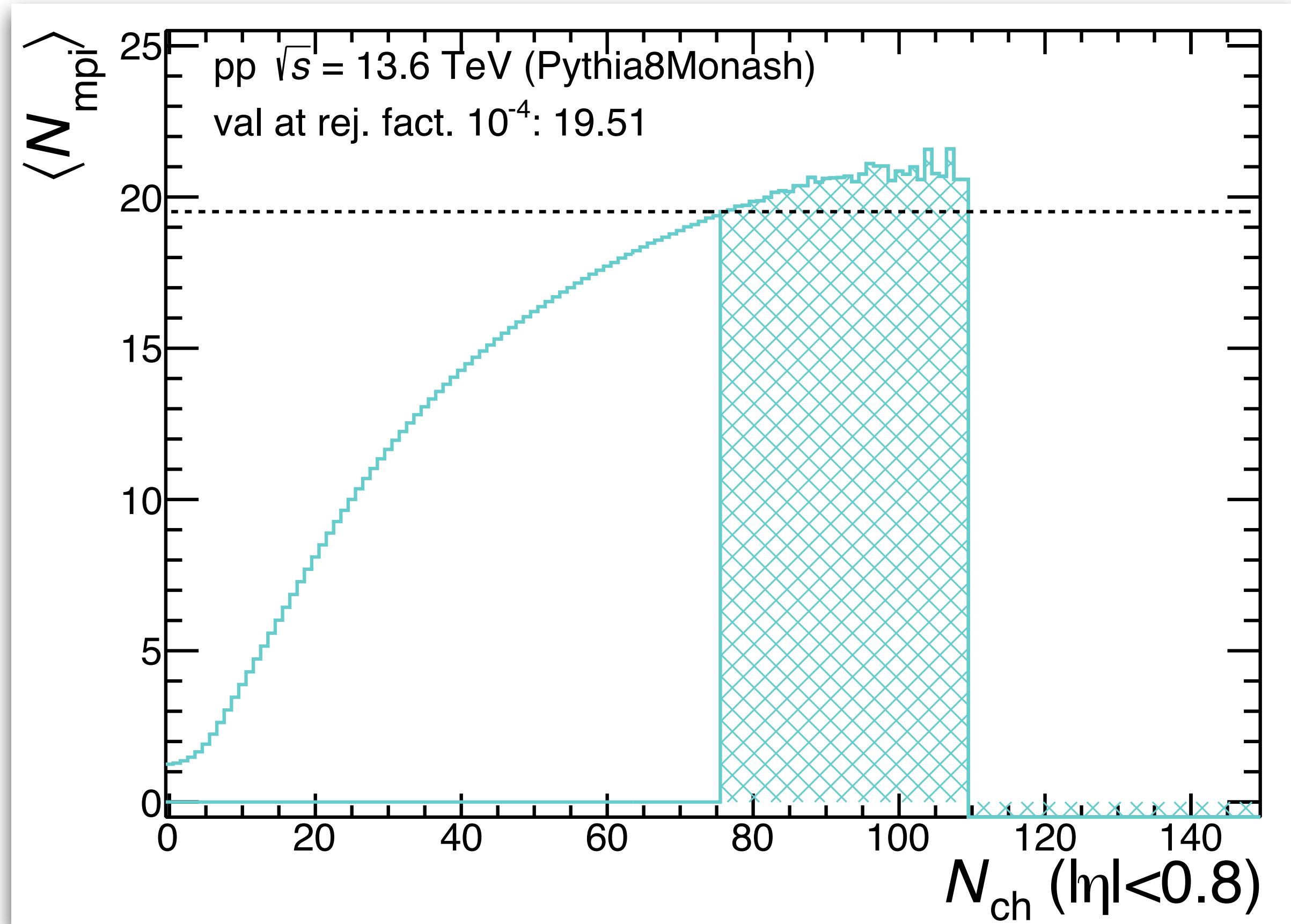


$$R_{\text{pp}} = \frac{d^2N_{\text{ch}}^{\text{mpi}} / (\langle N_{\text{mpi}} \rangle d\eta dp_T)}{d^2N_{\text{ch}}^{\text{MB}} / (\langle N_{\text{mpi}}^{\text{MB}} \rangle d\eta dp_T)}$$

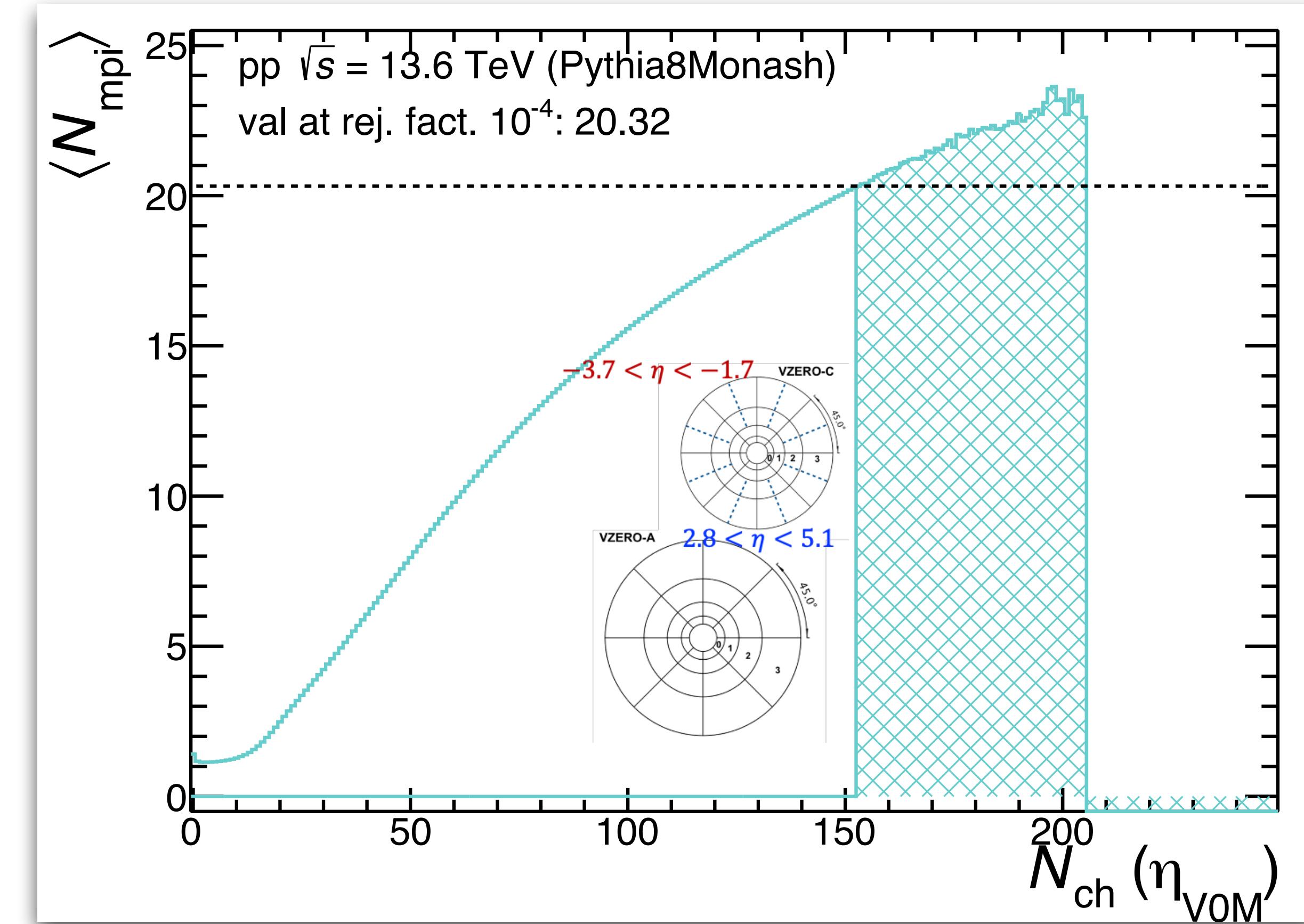
[Phys. Rev. D 102 \(2020\) 7, 076014](#)

# How can we tag those pp collisions?

Particle multiplicity seems to be a good candidate, example:



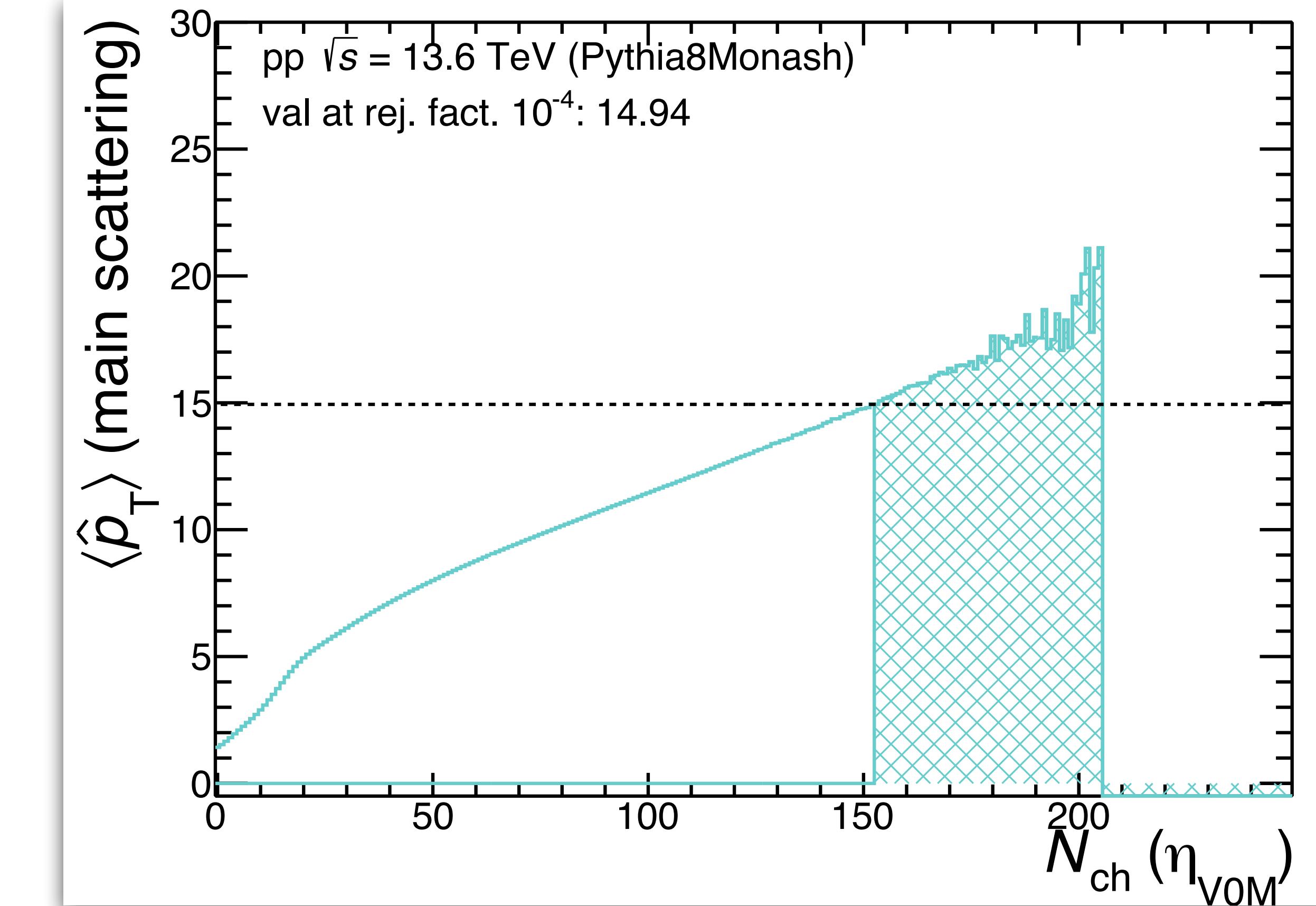
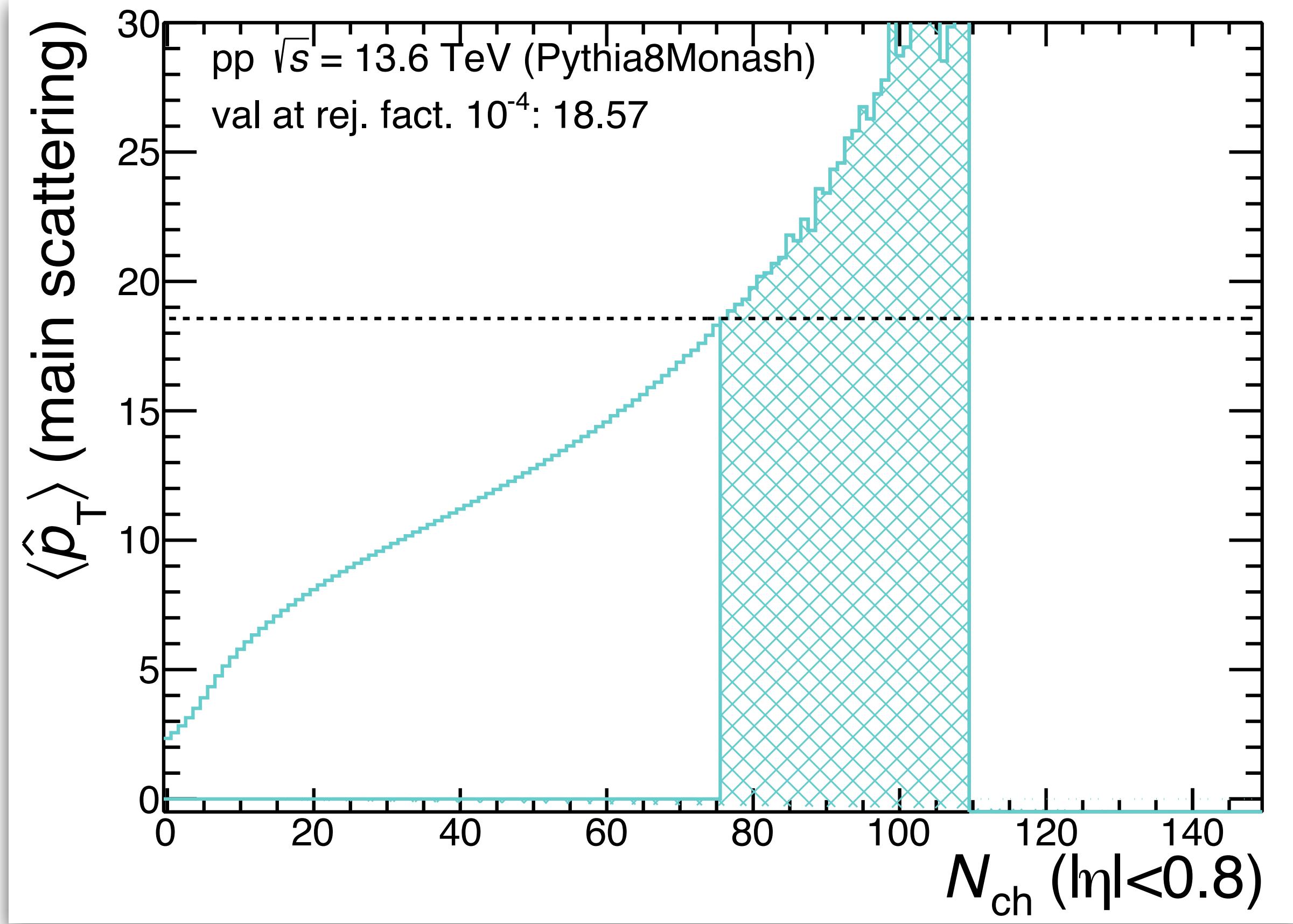
Track multiplicity ( $|\eta| < 0.8$ )



Multiplicity in the forward direction (VOM):  
 $-3.7 < \eta < -1.7$  and  $2.8 < \eta < 5.1$

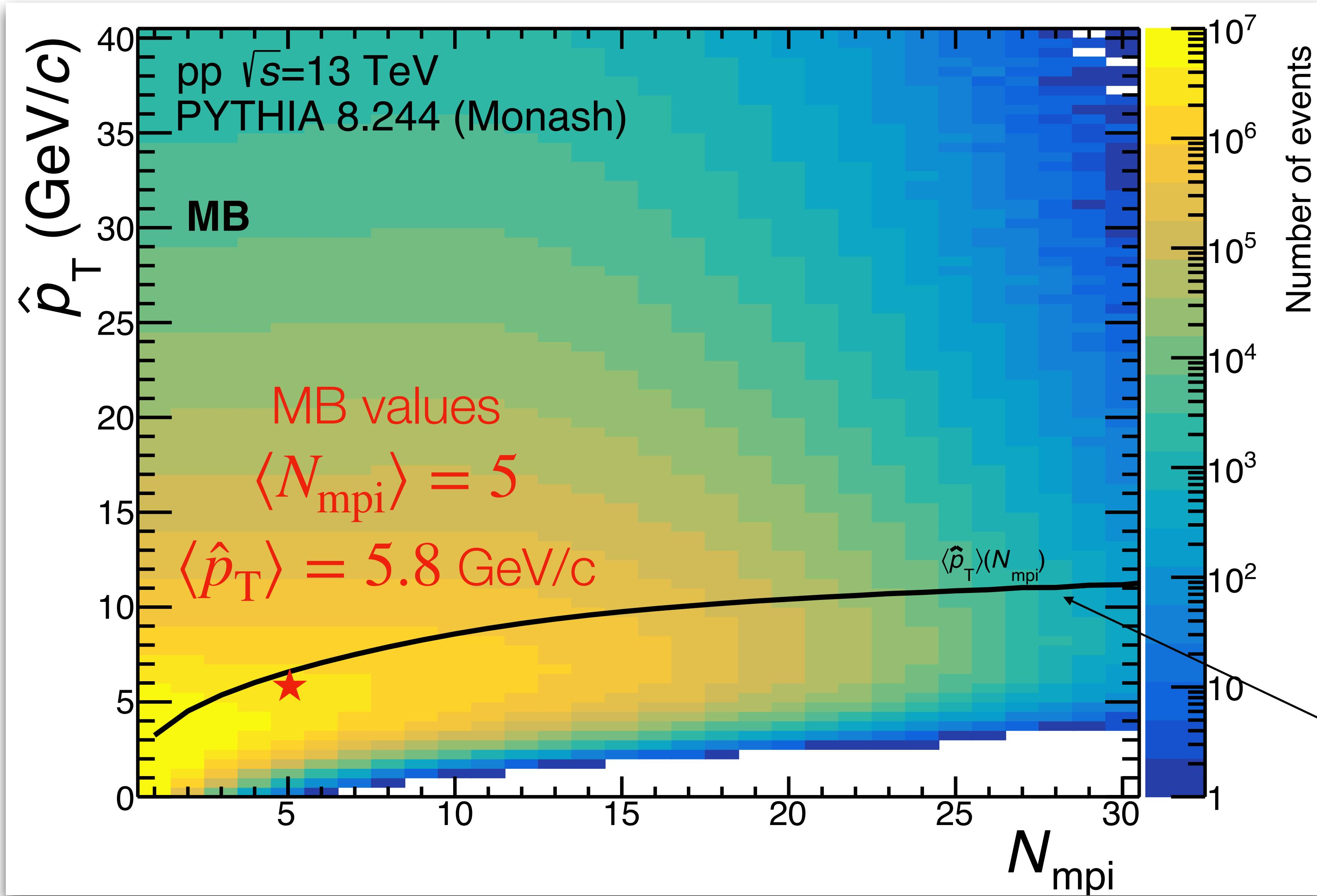
# Do we understand the bias of the different classifiers?

Example: strong sensitivity to MPI but bias towards hard pp collisions. The effect is stronger for the mid-pseudorapidity multiplicity estimator



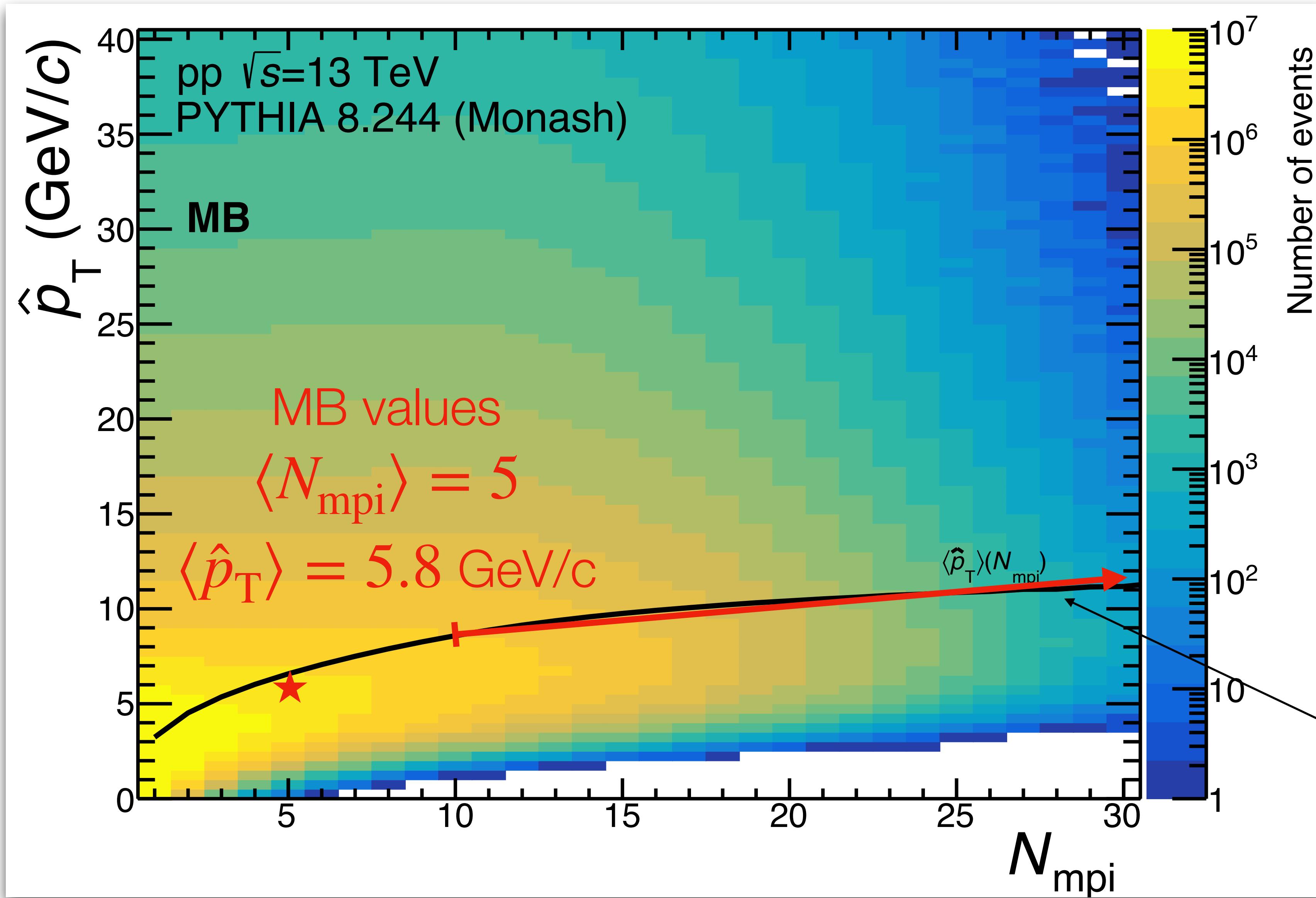
VOM seems to be more convenient, in addition no bias in the ratio charged-to-neutral particles

# Understanding the different event classifiers

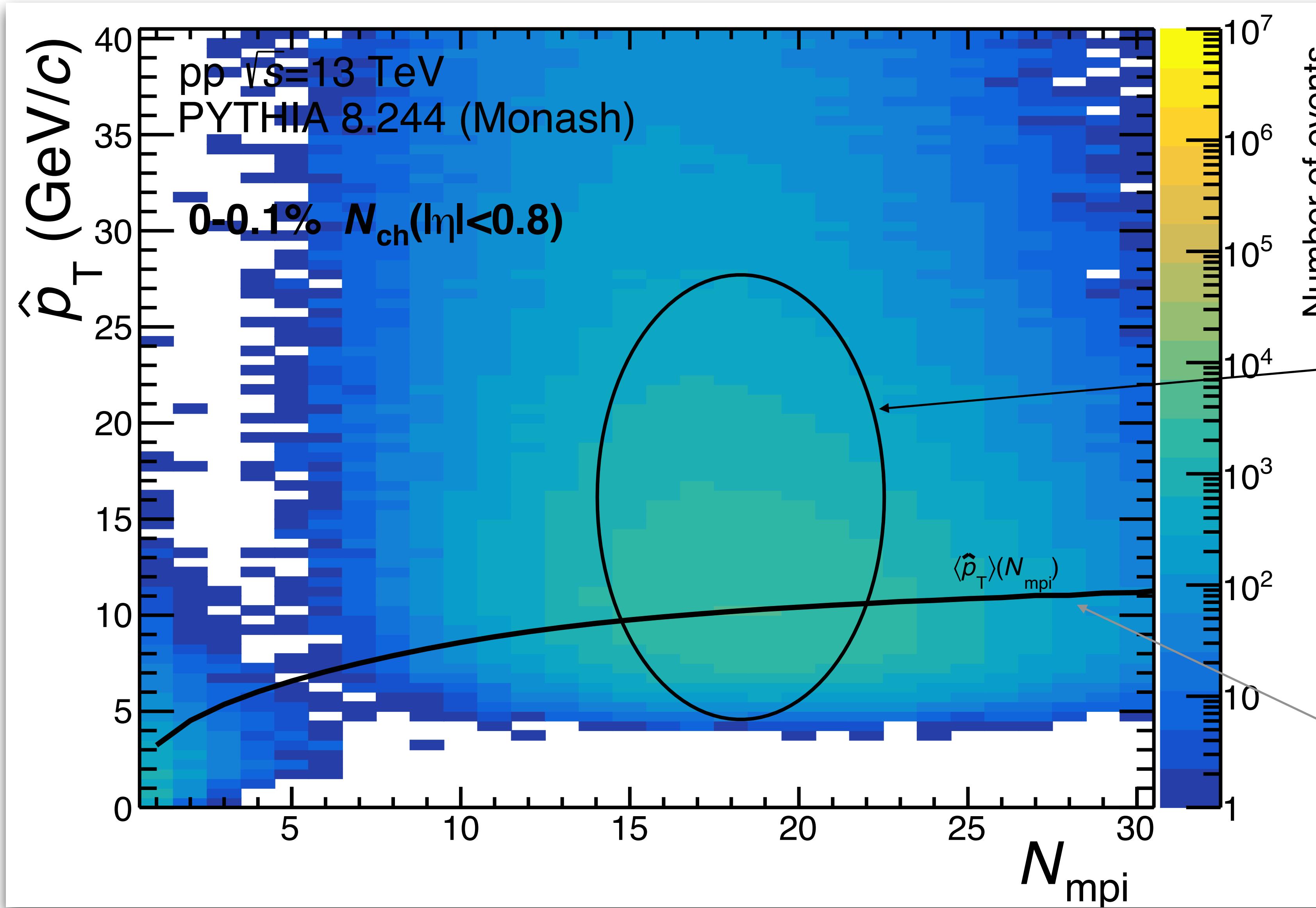


MB pp collisions:  
The solid line represents  
the unbiased correlation  
(our goal) between jet  $p_T$   
and  $N_{\text{mpi}}$

# Understanding the different event classifiers



# High-multiplicity selector $N_{\text{ch}}$

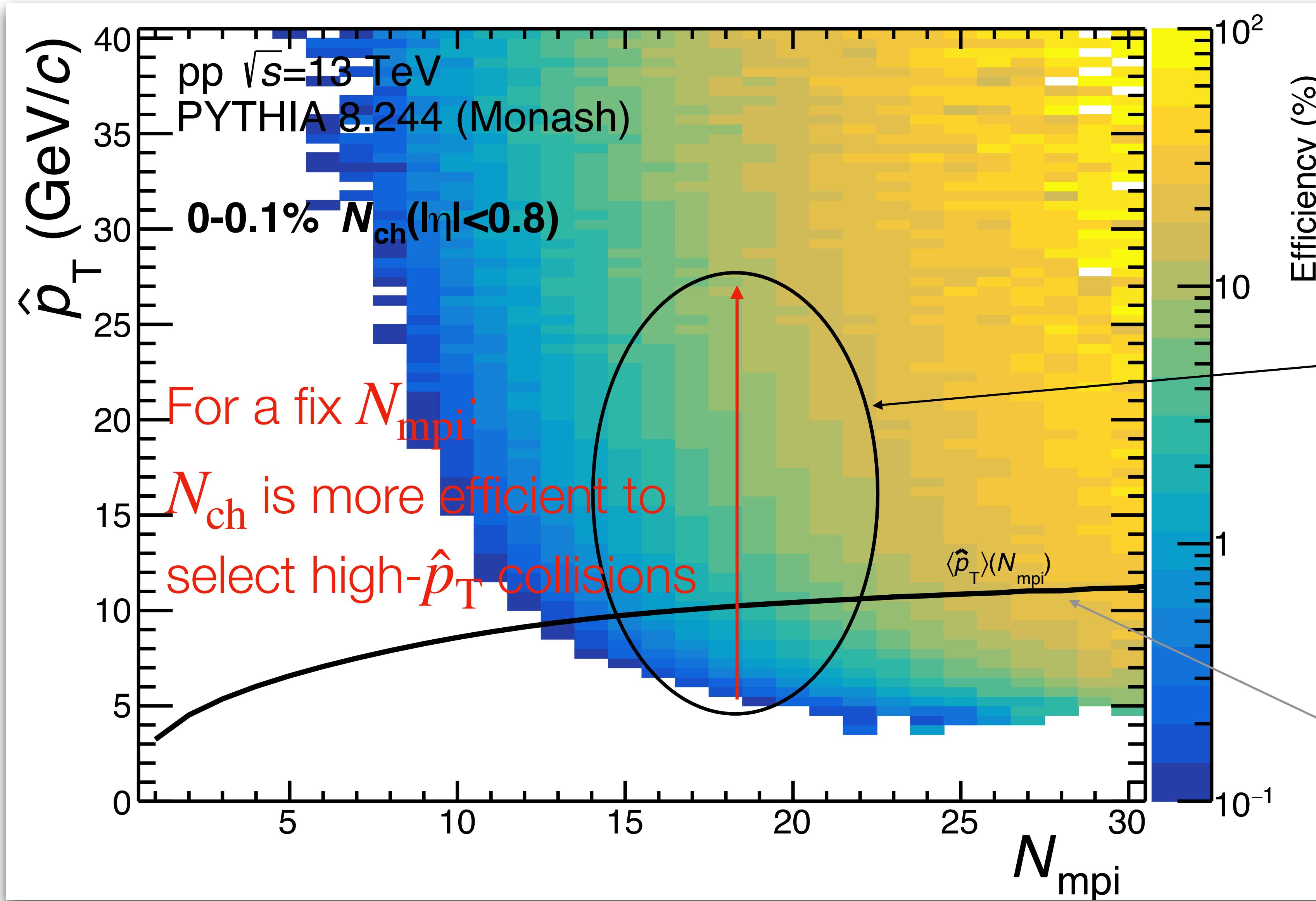


0-0.1%  $N_{\text{ch}}$ :  
Primary charged particles  
 $|\eta| < 0.8, p_T > 0 \text{ GeV}/c$

$\sigma_{\hat{p}_T} - \sigma_{N_{\text{mpi}}}$ :  
Correlation ellipse

MB pp collisions:  
The solid line represents  
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# High-multiplicity selector (efficiency)

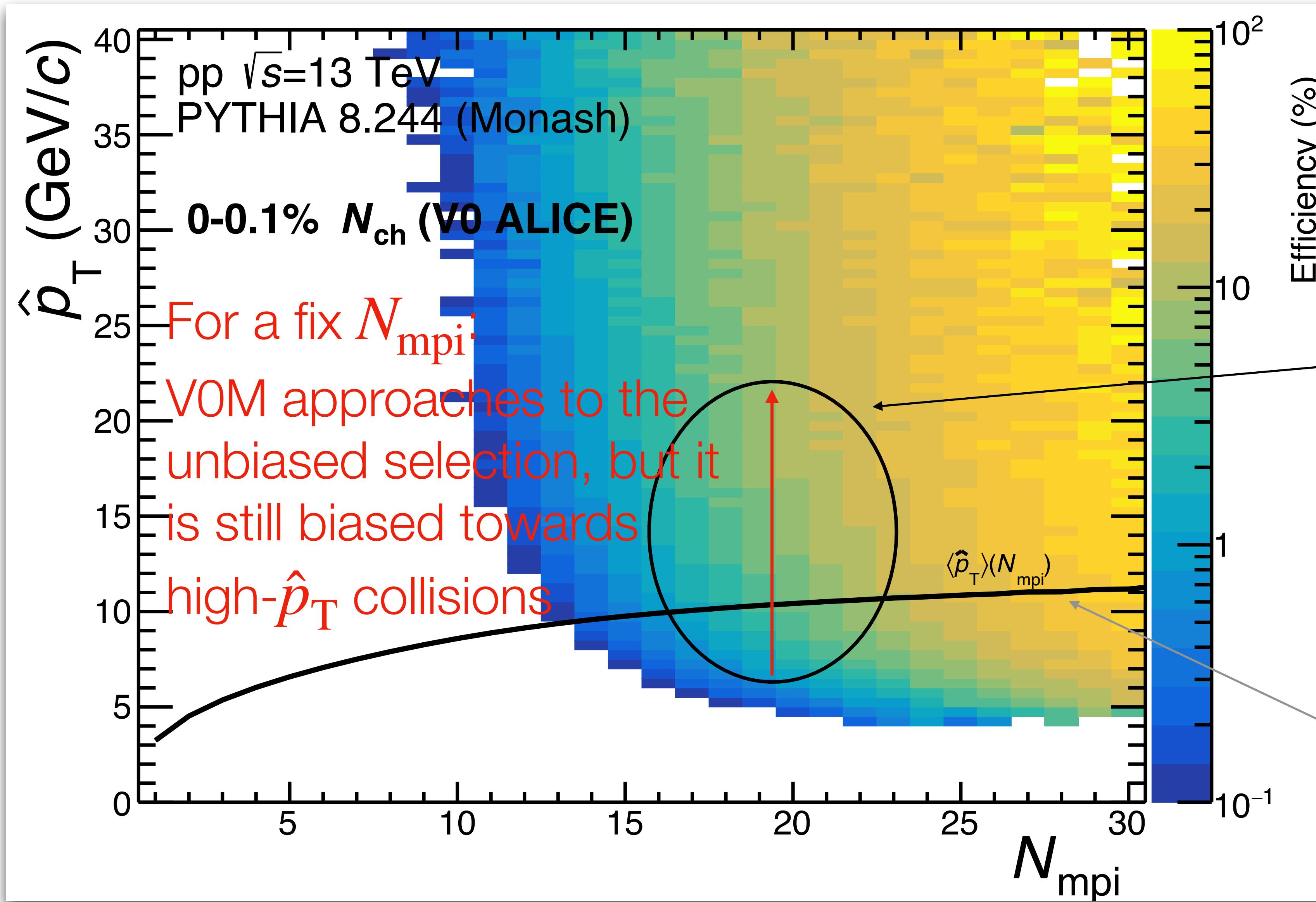


0-0.1%  $N_{\text{ch}}$ : Primary charged particles  $|\eta| < 0.8, p_T > 0 \text{ GeV}/c$

$\sigma_{\hat{p}_T} - \sigma_{N_{\text{mpi}}}$ : Correlation ellipse

MB pp collisions:  
The solid line represents the unbiased correlation (our goal) between jet  $p_T$  and  $N_{\text{mpi}}$

# Efficiency: $N_{\text{ch}} \rightarrow \text{V0M}$

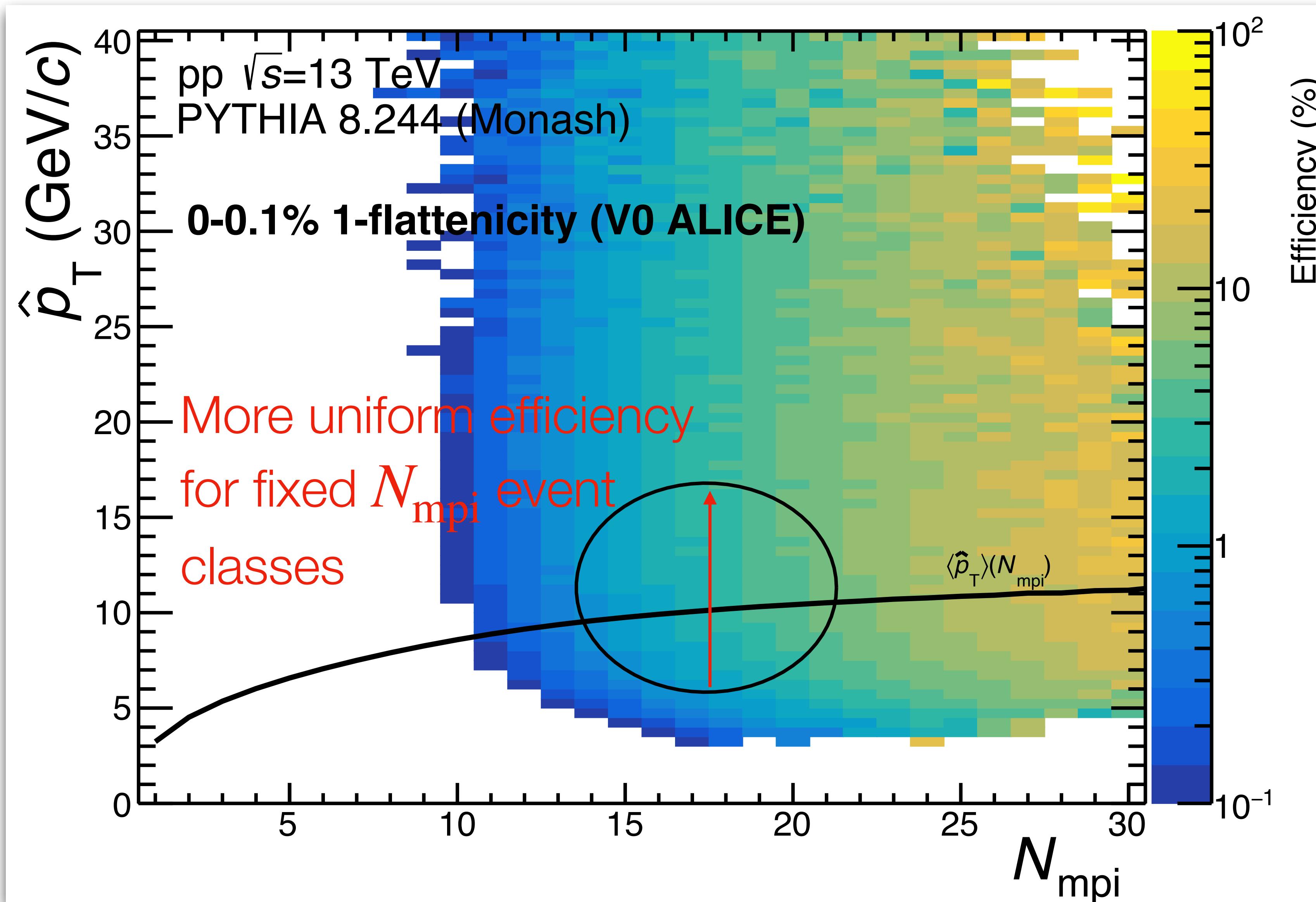


0-0.1% V0M:  
Primary charged particles ( $p_T > 0 \text{ GeV}/c$ ) in the ALICE V0 acceptance

$\sigma_{\hat{p}_T} - \sigma_{N_{\text{mpi}}}$ :  
Correlation ellipse (narrower area than with  $N_{\text{ch}}$ )

MB pp collisions:  
The solid line represents the unbiased correlation (our goal) between jet  $p_T$  and  $N_{\text{mpi}}$

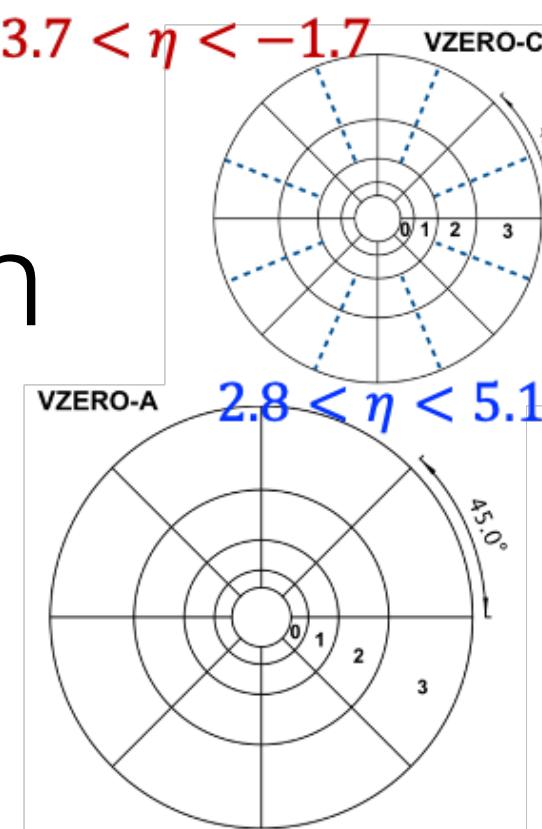
# Efficiency: $N_{\text{ch}} \rightarrow \text{V0M} \rightarrow \text{flattenicity}$



Calculated in the region cover by the V0 detector and using  $N_{\text{ch}}$  instead of  $p_T$

$$\rho = \frac{\sqrt{\sum_i^{N_{\text{cell}}} (N_{\text{ch}}^i - \langle N_{\text{ch}} \rangle)^2 / N_{\text{cell}}^2}}{\langle N_{\text{ch}} \rangle}$$

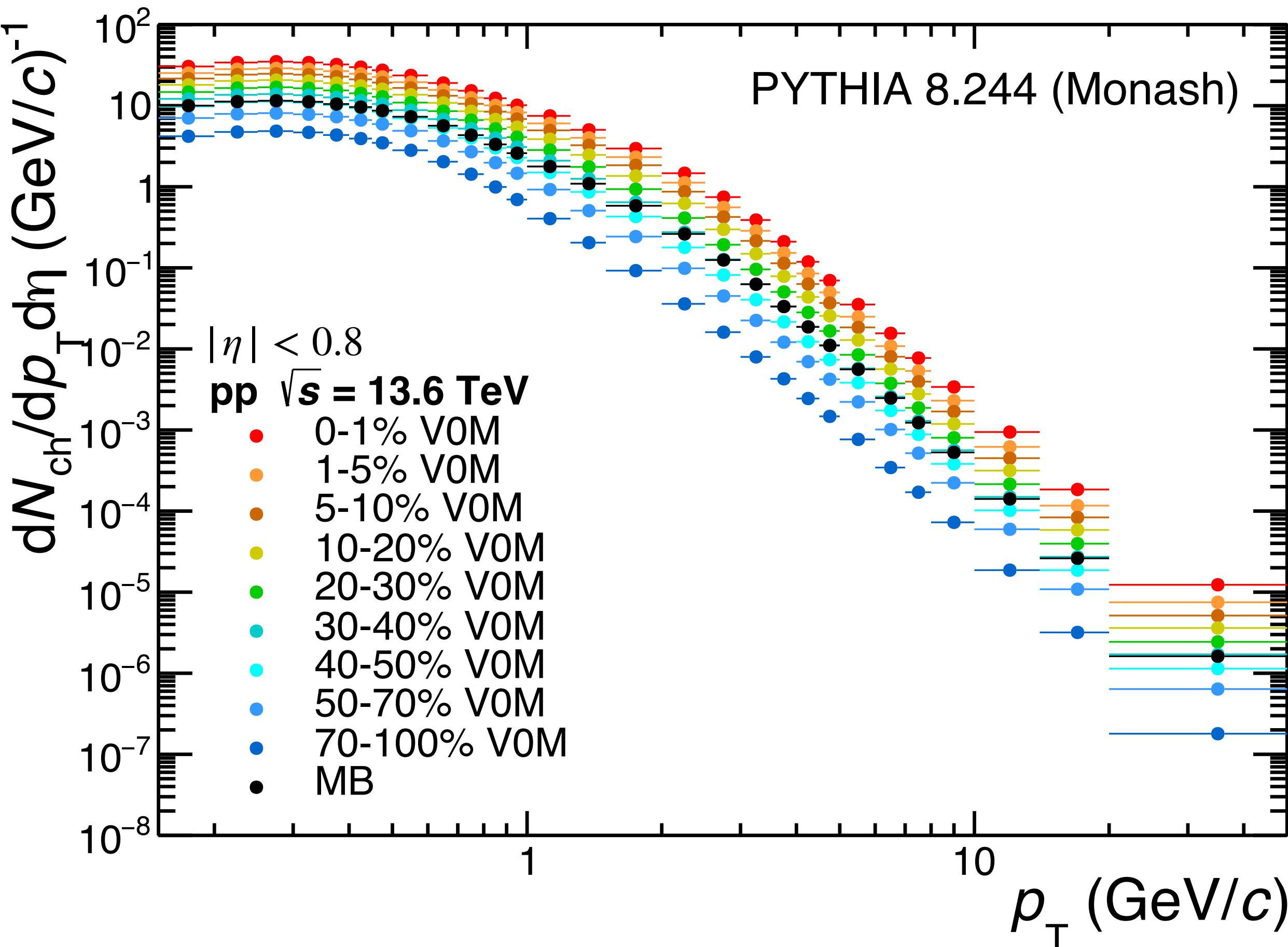
The normalisation to  $N_{\text{cell}}$  allows flattenicity to stay within 0-1



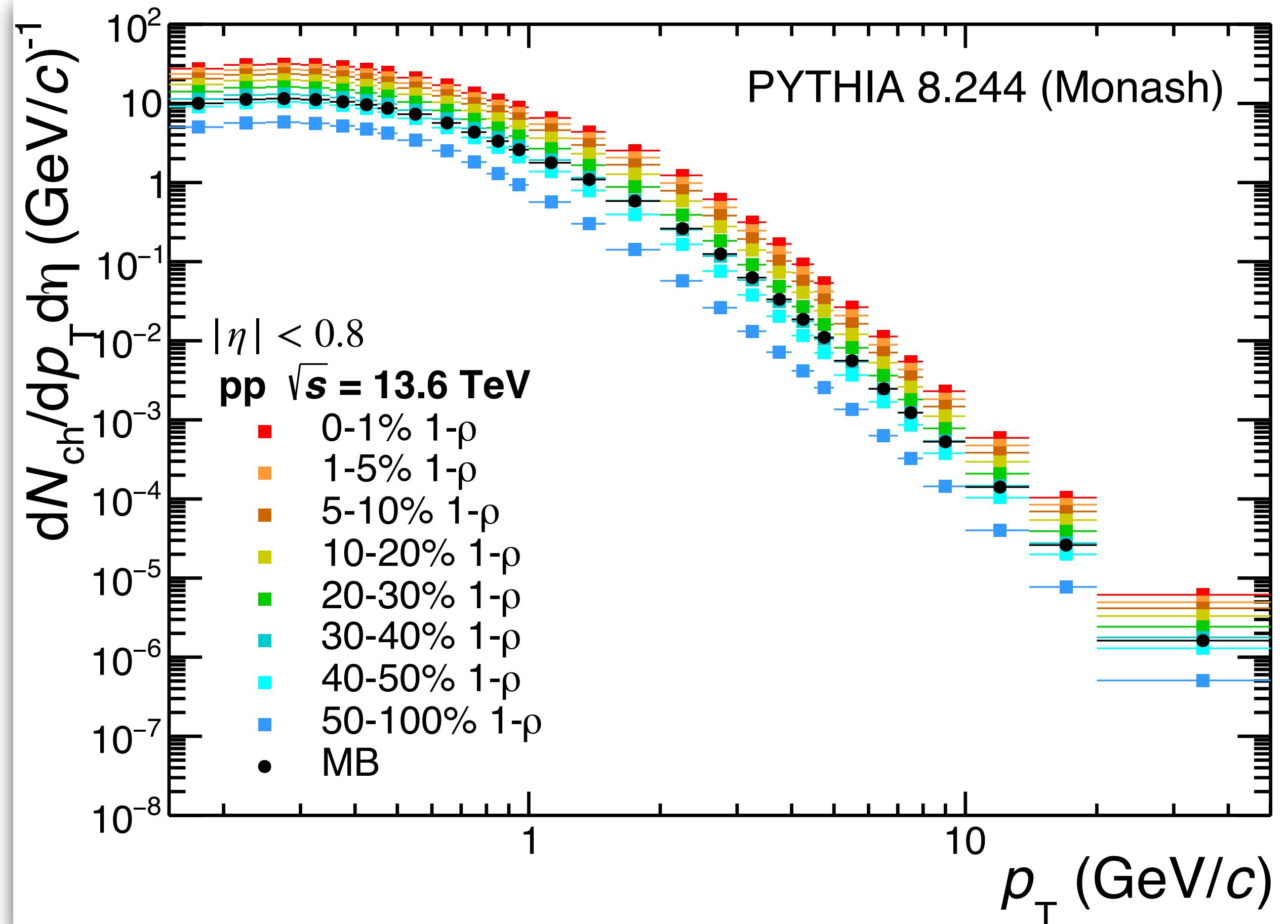
[A. Ortiz et al., PRD 107 \(2023\) 7, 076012](#)

# Biases at work! VOM vs flattenicity selections

Selection based on VOM

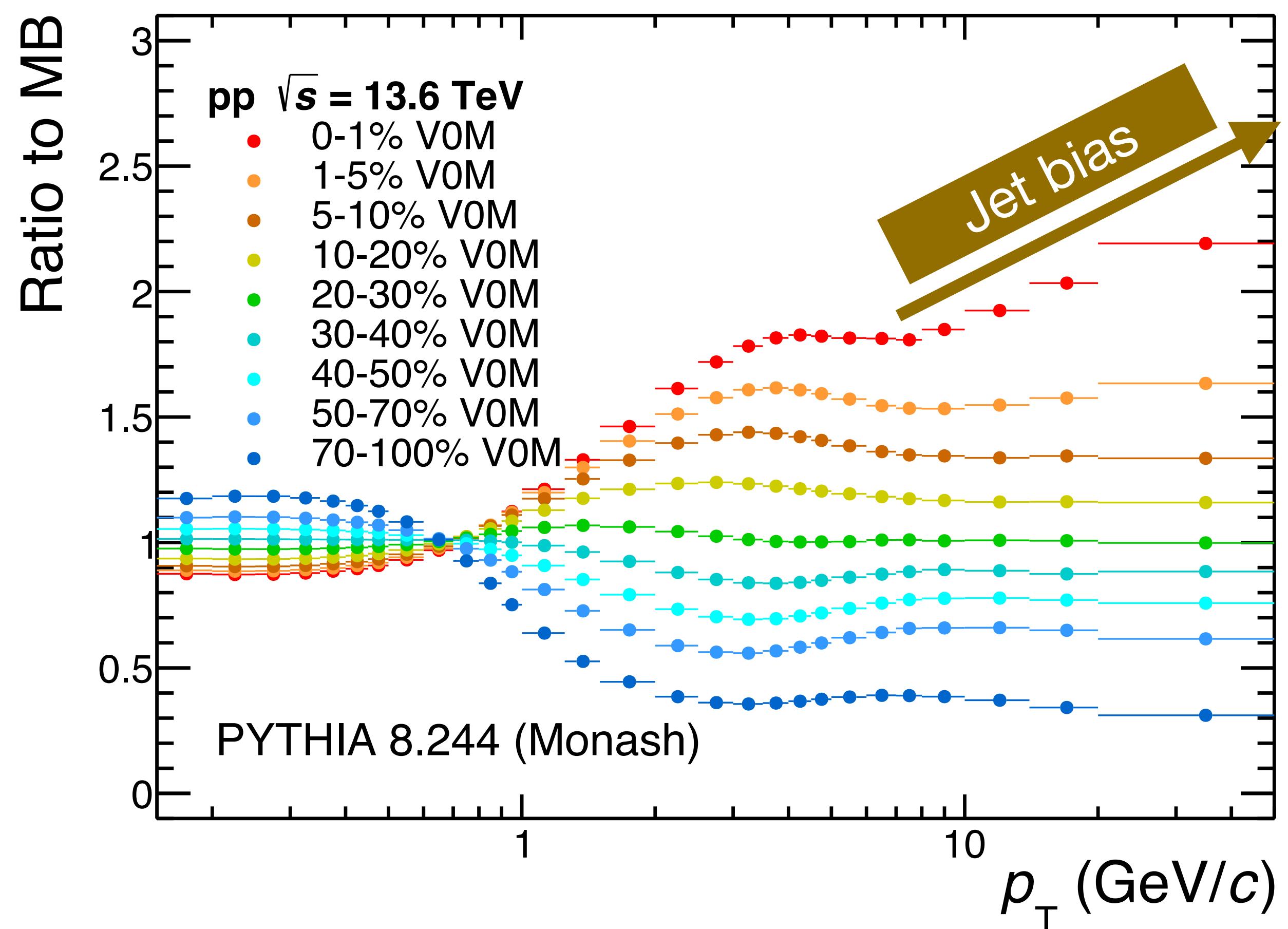


Selection based on flattenicity

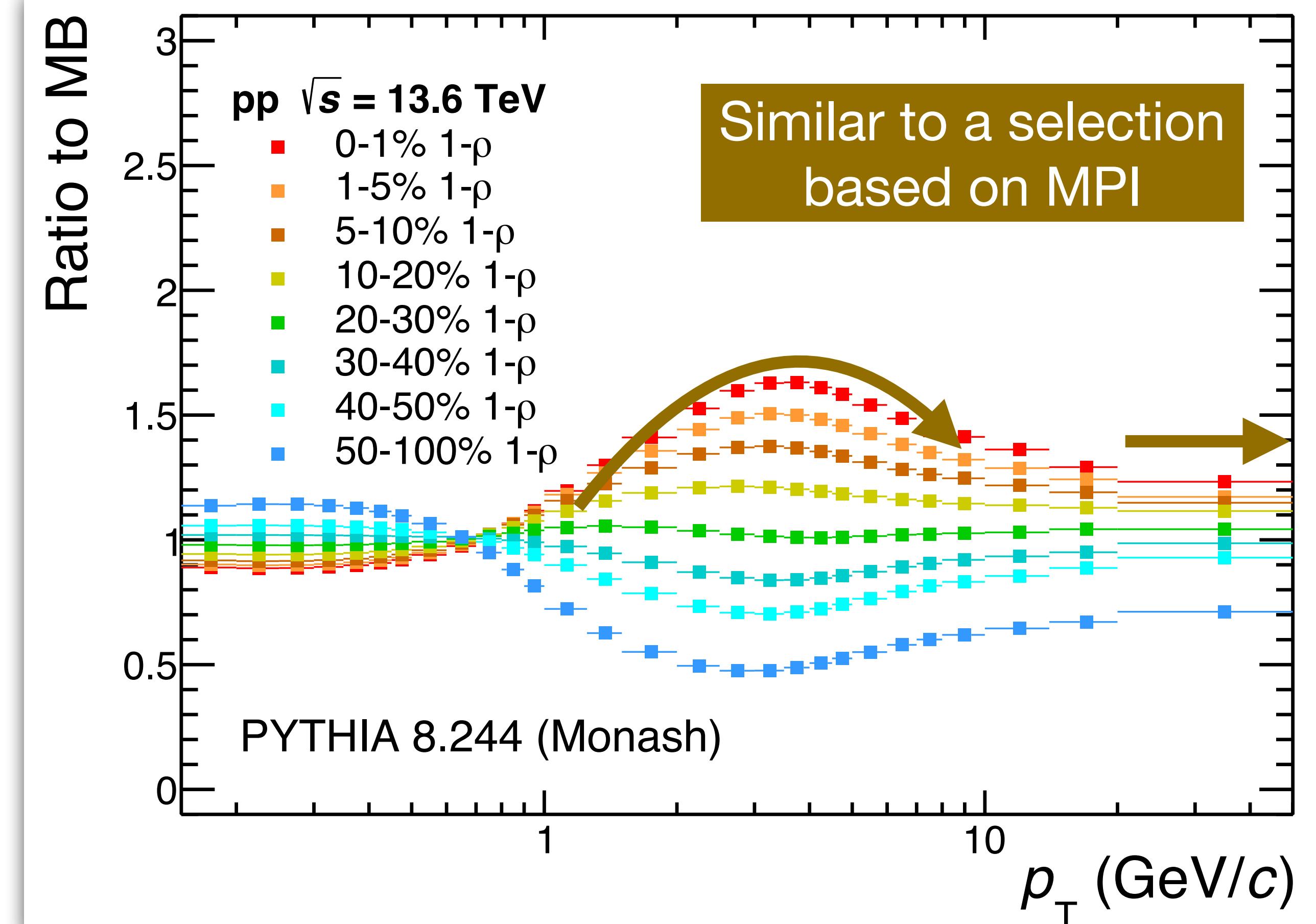


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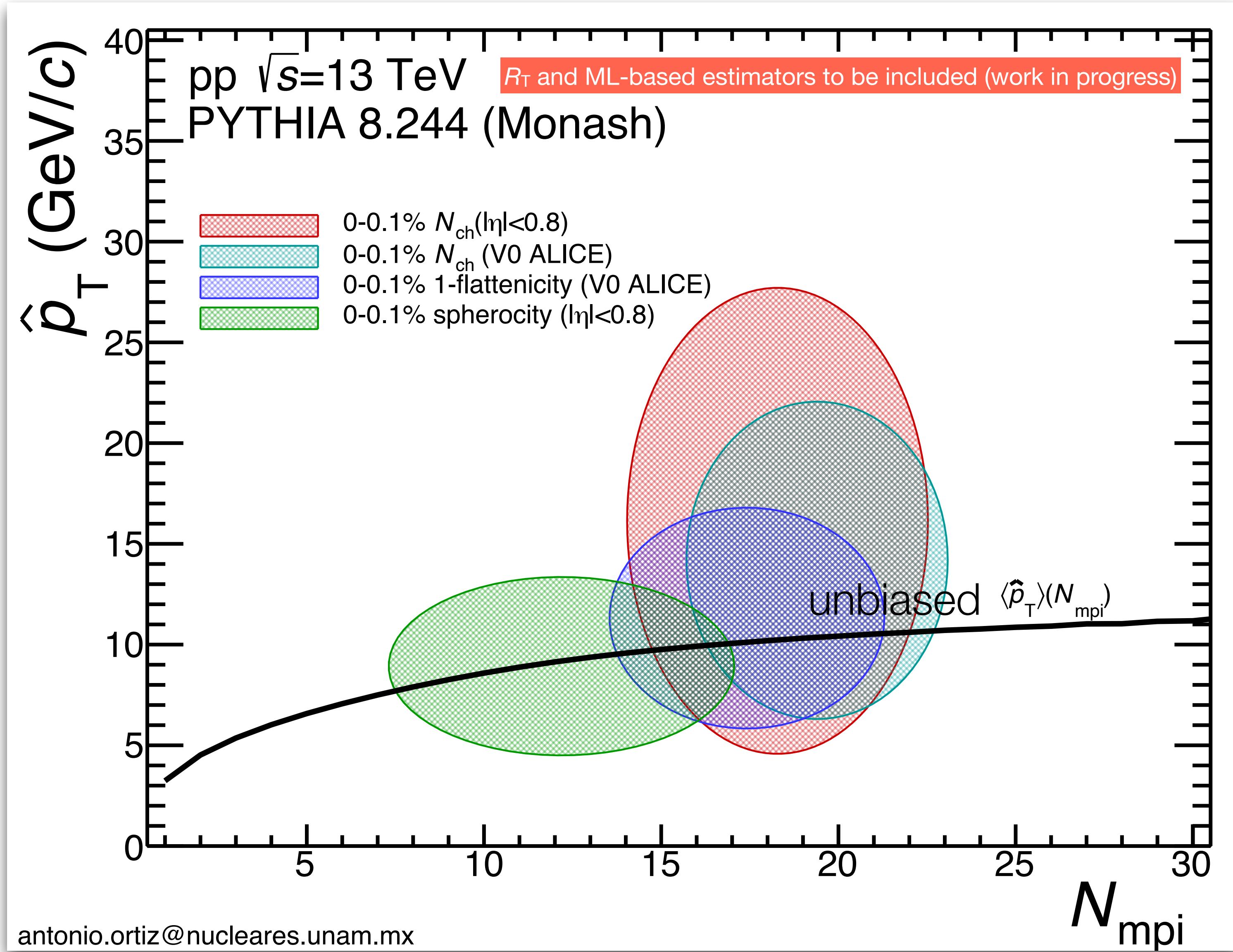
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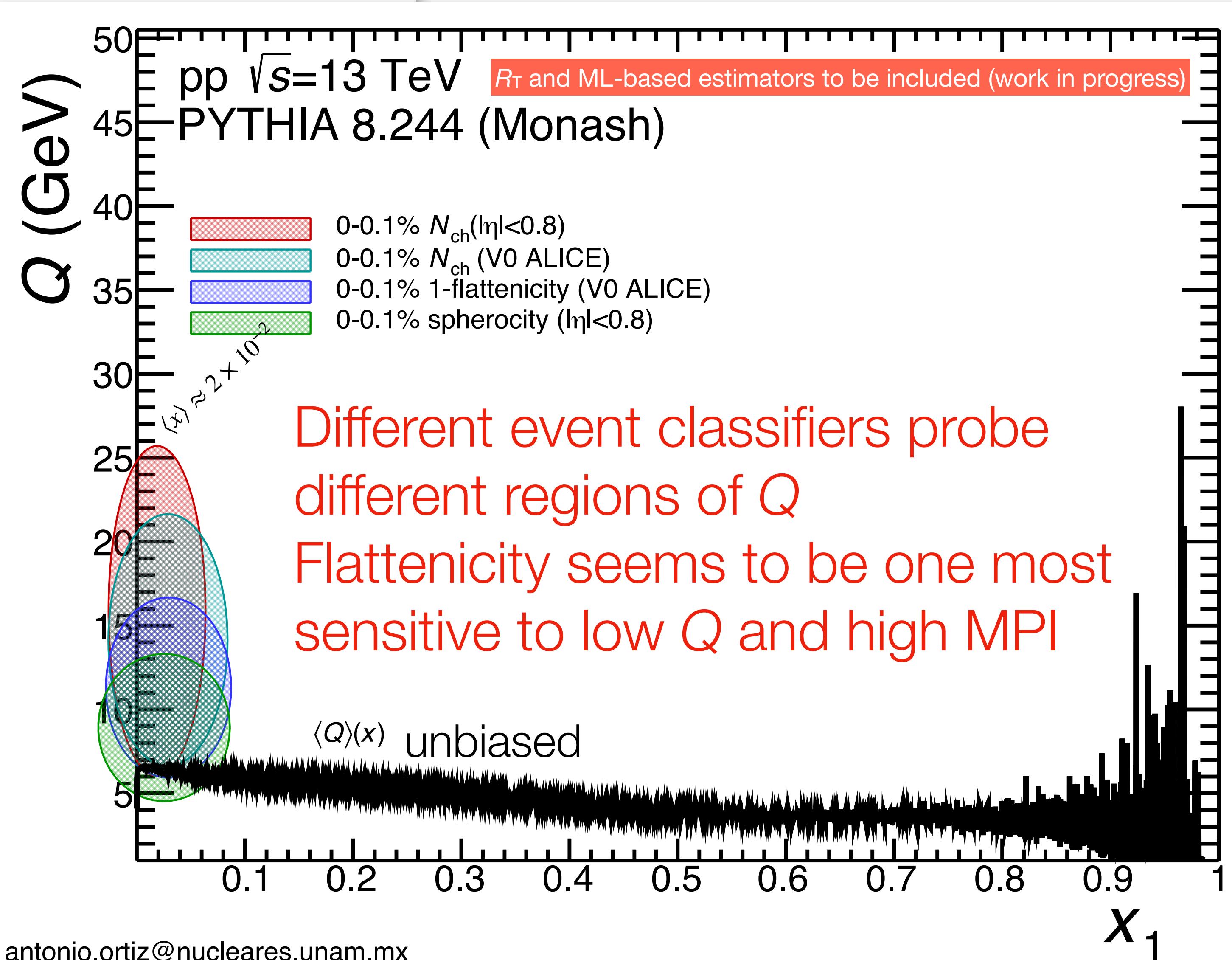
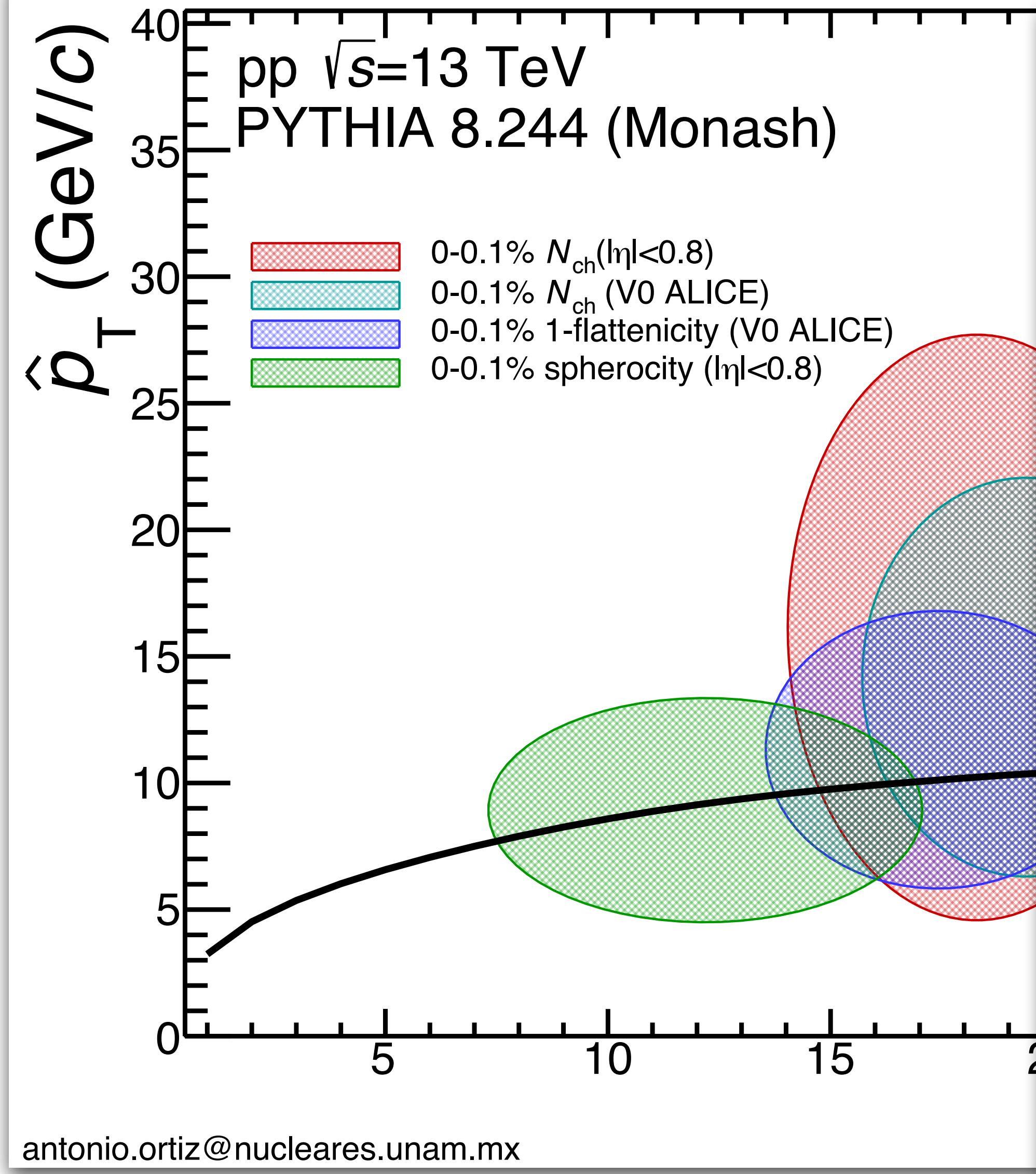
Selection based on flattenicity



# How are the different classifiers related to each other?



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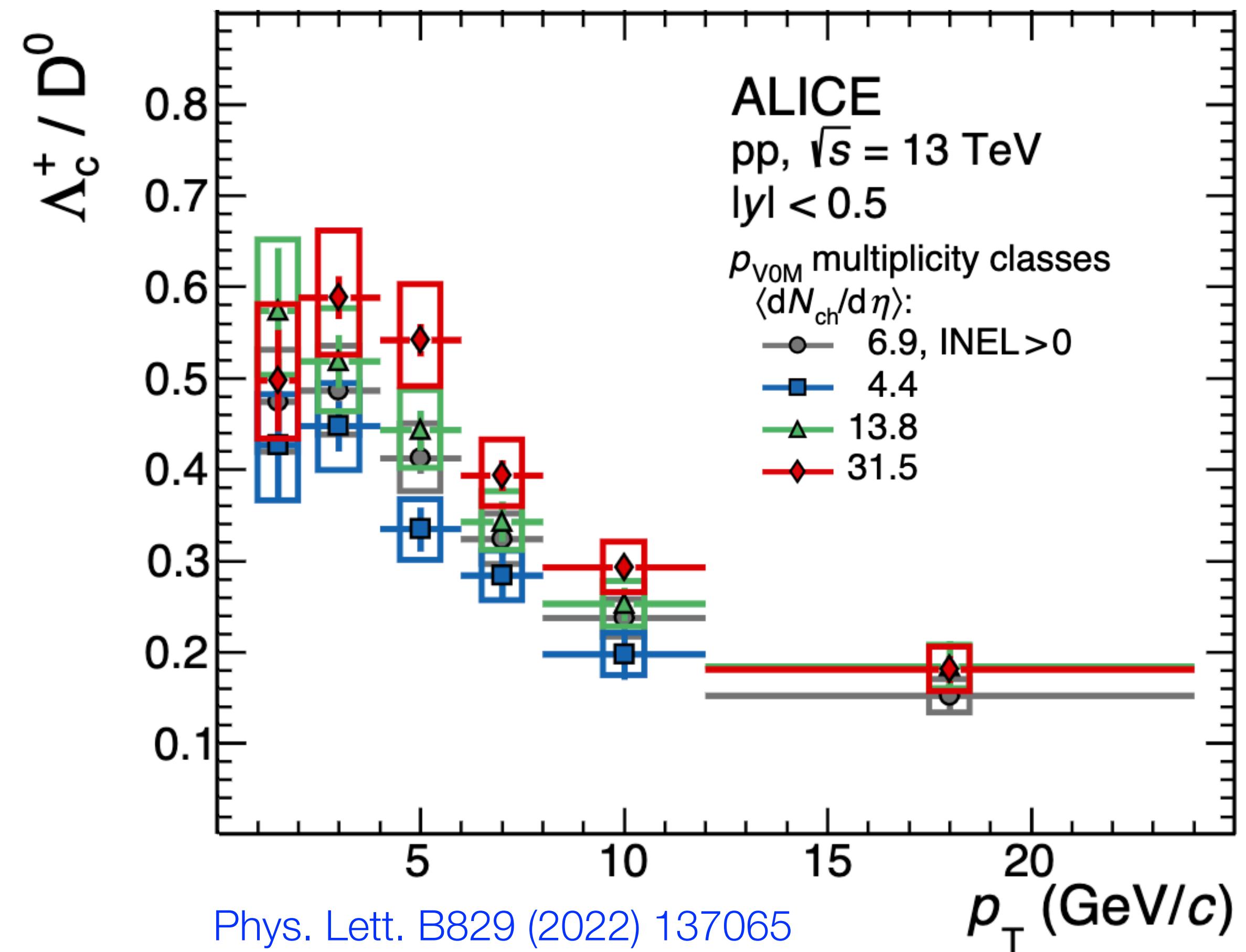


# Take home message

- What type of pp collisions should we compare with heavy-ion collisions (HIC)?  
**Events with large MPI activity but with “jet bias” well under control**
- How can we tag those pp collisions (event classifiers)  
**Several options in the market, flattenicity seems like one of the most promising**
- Do we understand the bias of the different classifiers explored so far? How are they correlated?  
**MC studies like the one presented today need to be performed (see e.g. a study for  $R_T$ ; G. Bencedi, A. Ortiz, A. Paz, PRD 104 (2021) 1, 016017)**
- Are our data biased in the same way as MC?  
**Probably not at the same level, but this need to be understood**  
**Multi-experiment effort would be important**  
**Common strategy (perhaps one outcome of this workshop?)**

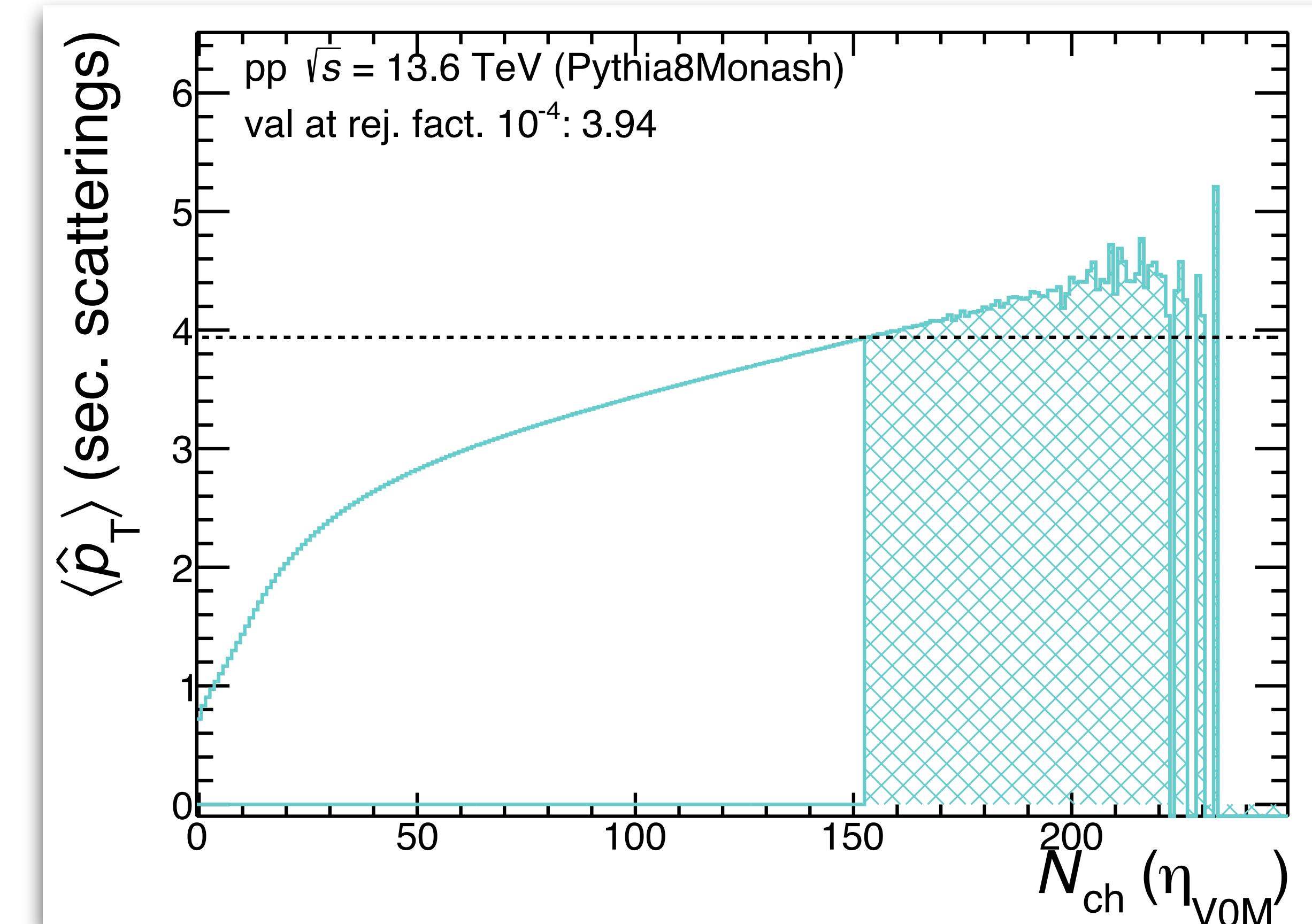
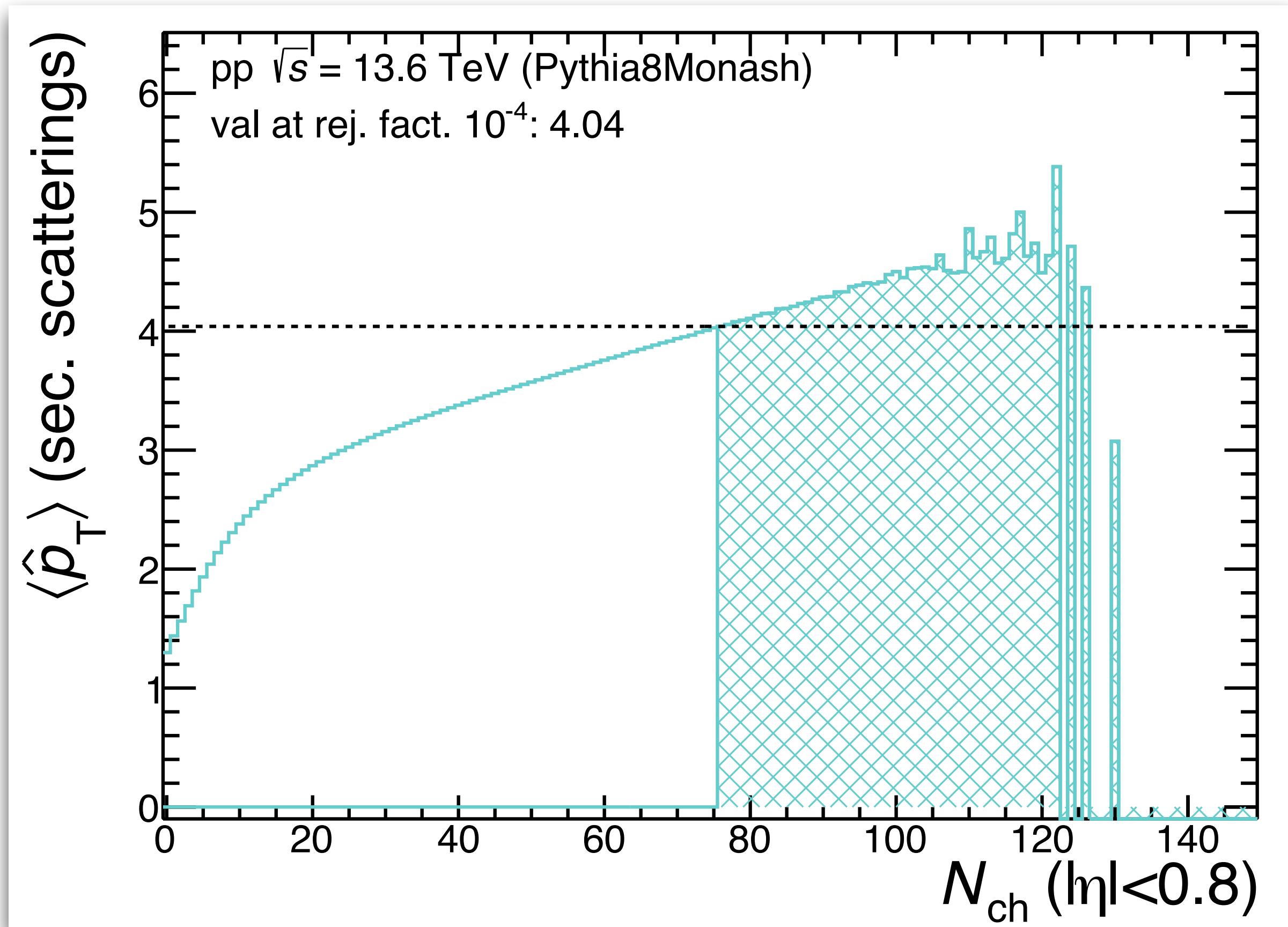
# Backup

# Particle ratios vs Nch

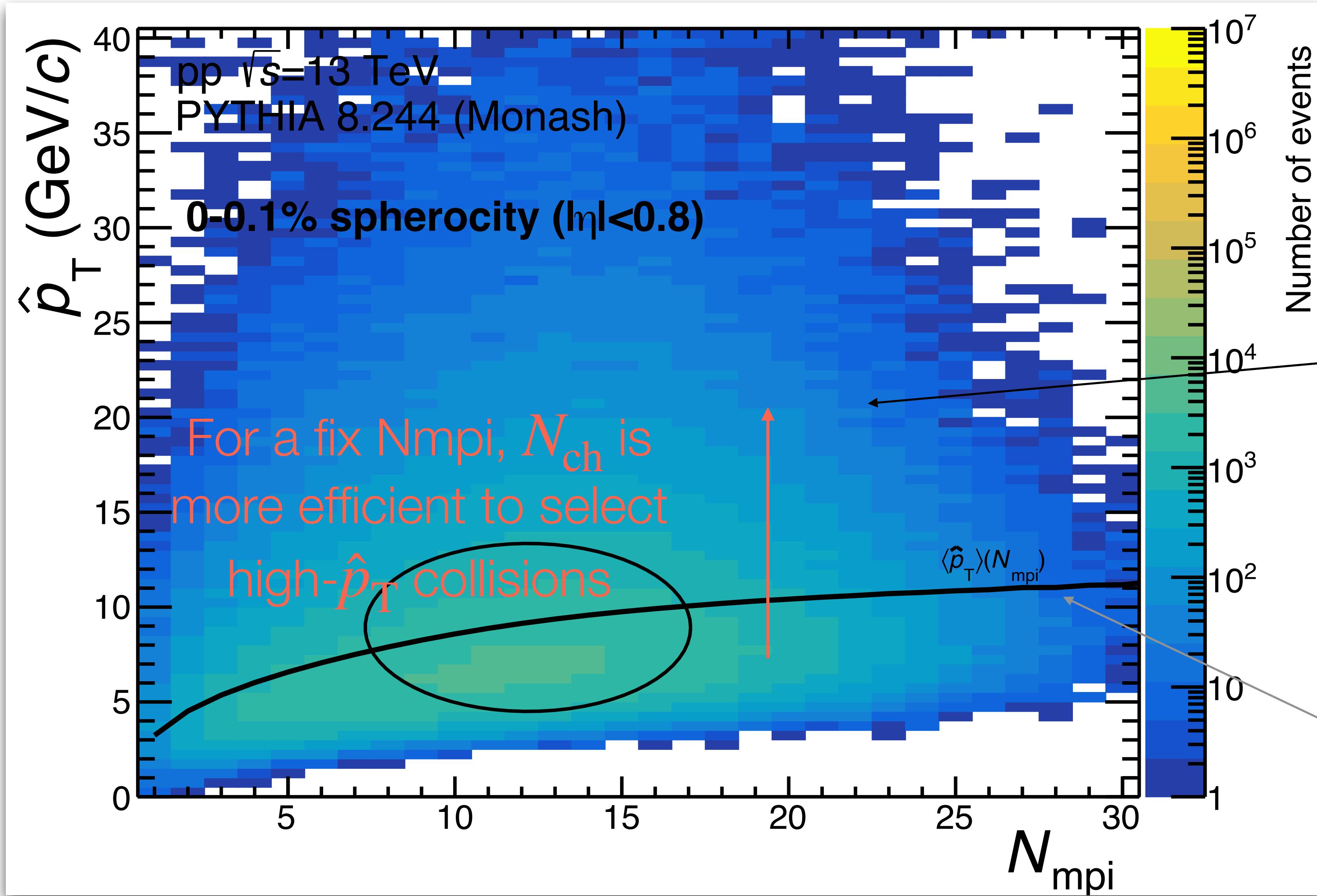


# Do we understand the bias of the different classifiers?

Example: strong sensitivity to MPI but bias towards hard pp collisions



# Spherocity selector (efficiency)



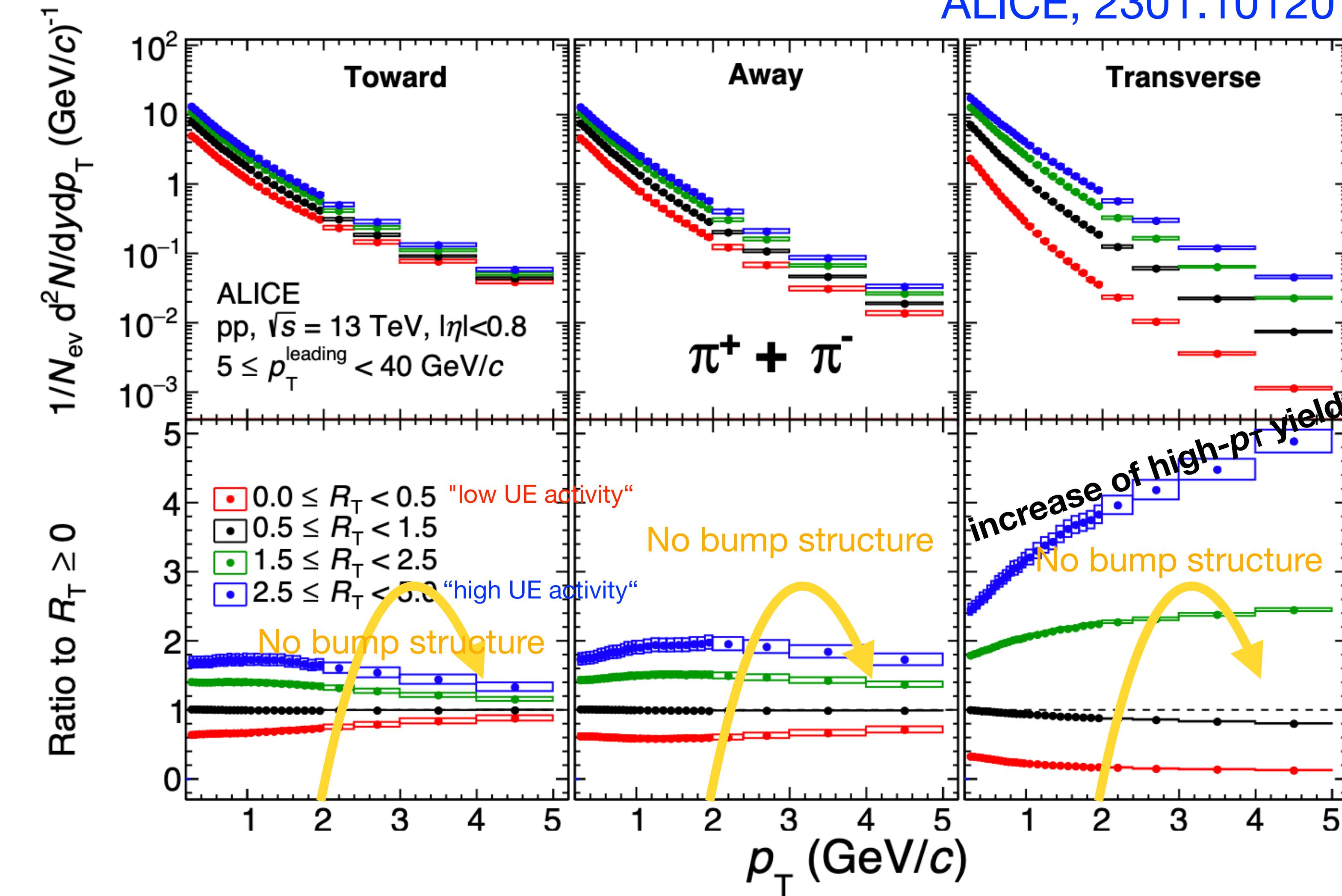
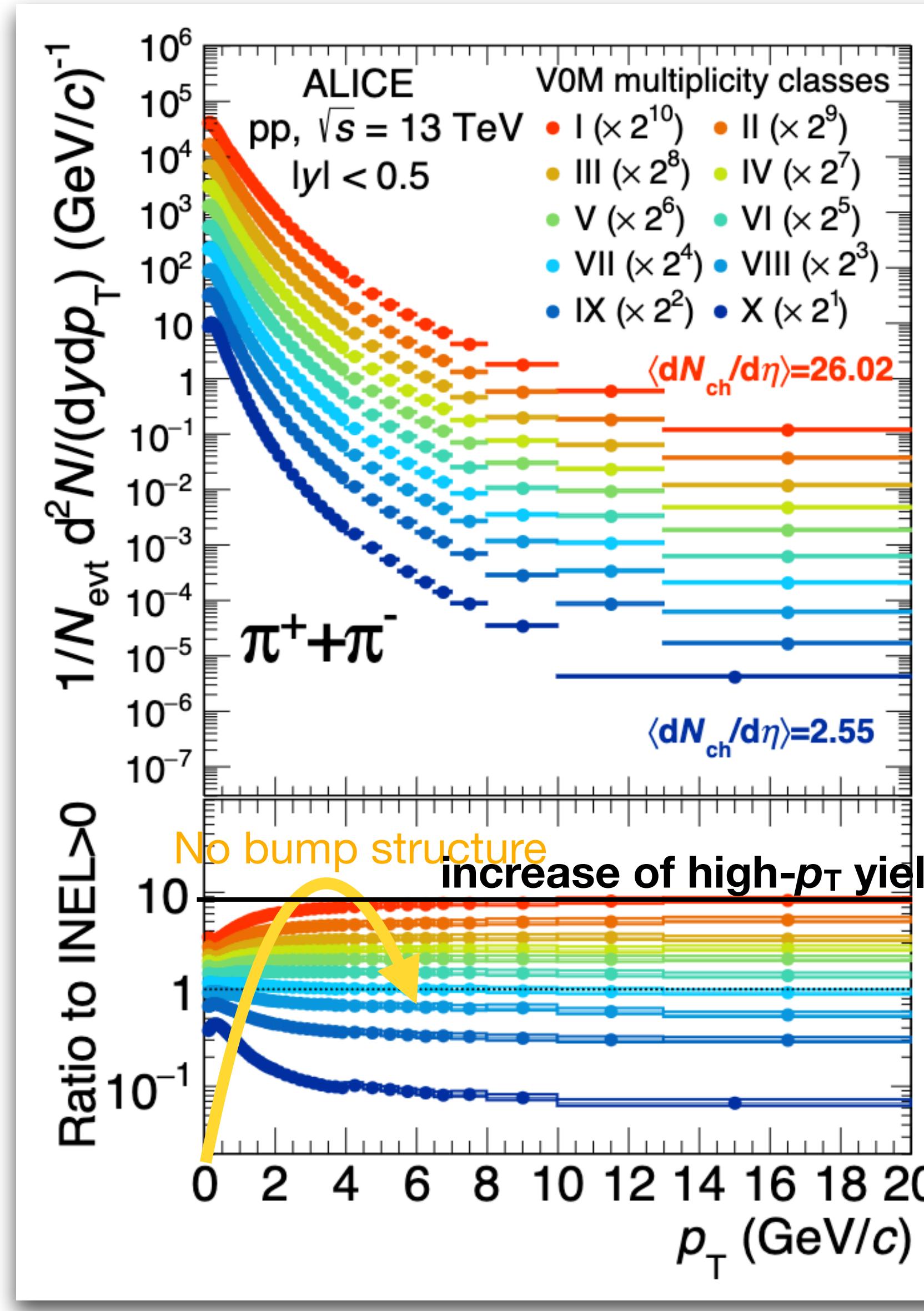
0-0.1%  $N_{\text{ch}}$ : Primary charged particles  $|\eta| < 0.8, p_T > 0$  GeV/c

$\sigma_{\hat{p}_T} - \sigma_{N_{\text{mpi}}}$ : Correlation ellipse

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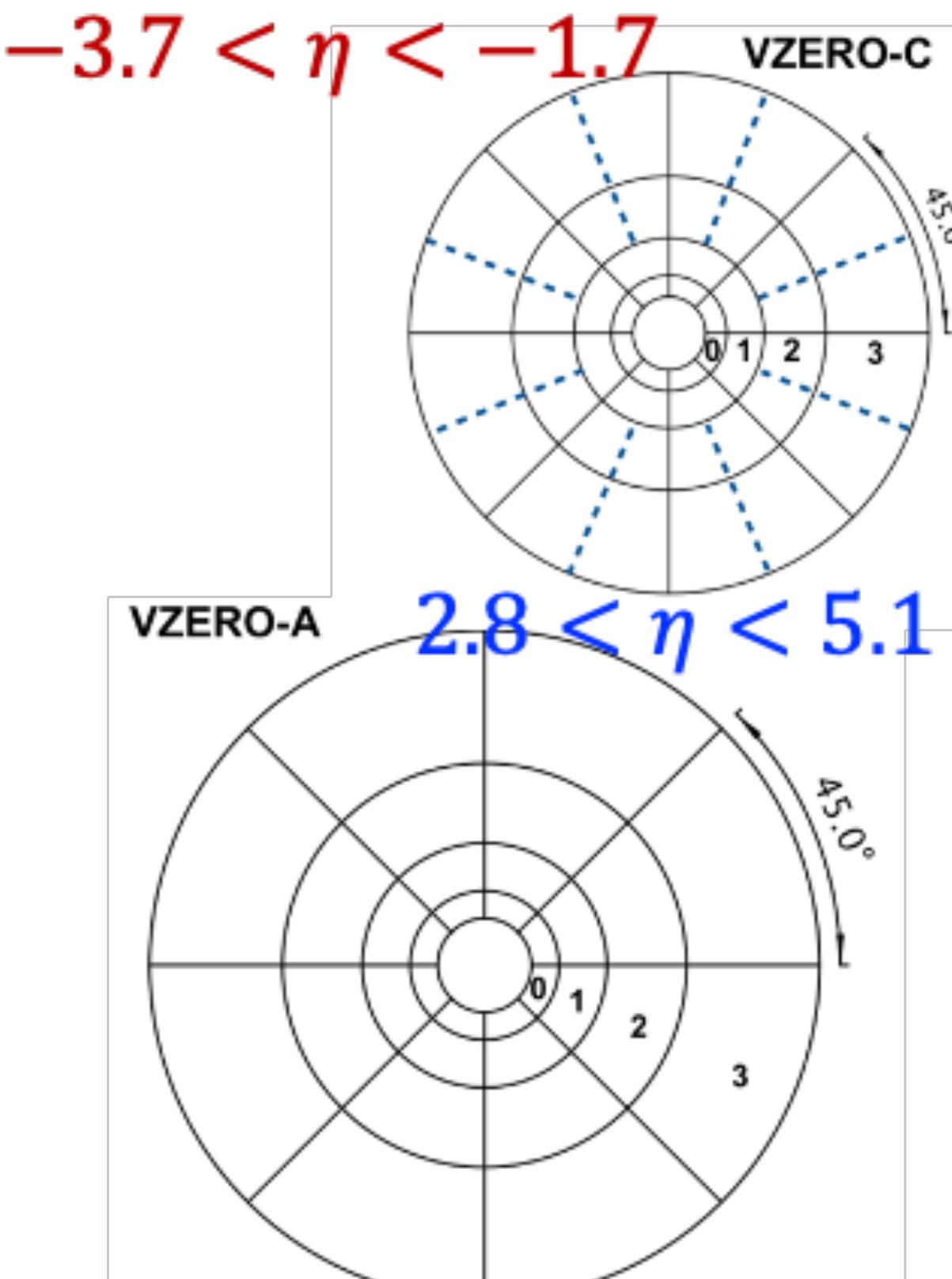
# Existing “ $R_{pp}$ ” results as a function of $N_{ch}$

ALICE, EPJC 80 (2020) 693

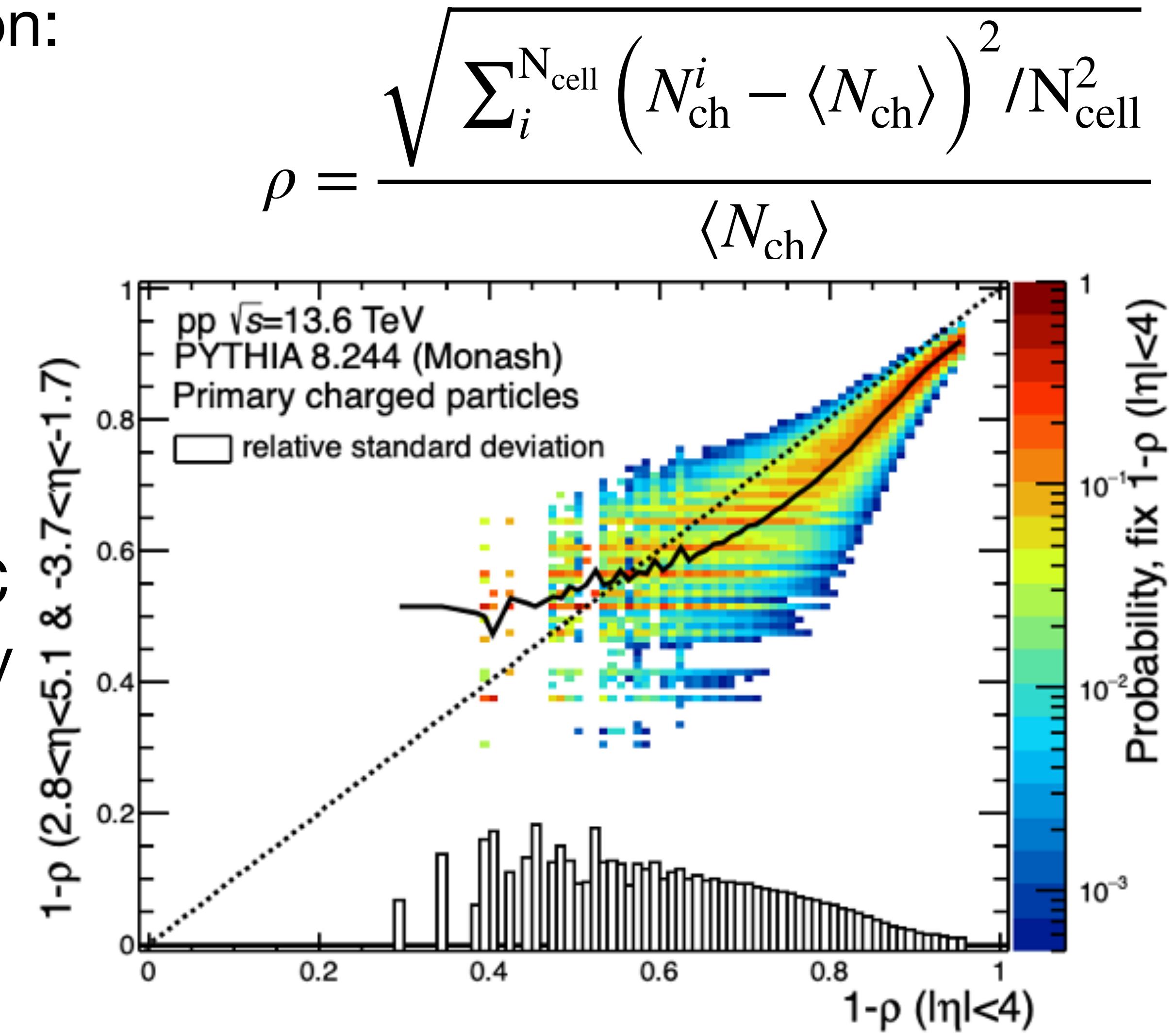


How can we increase the sensitivity to MPI?

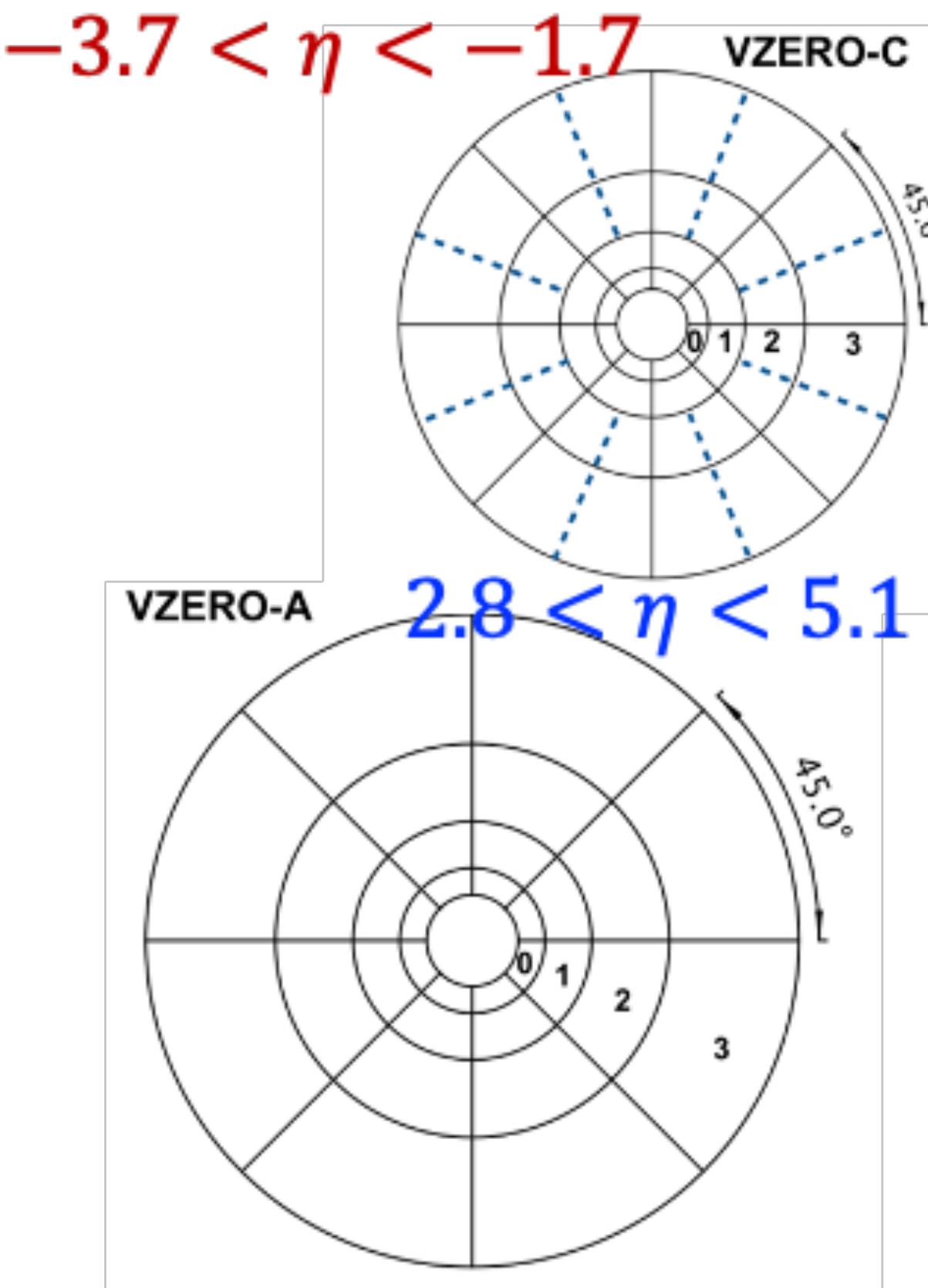
# Flattenicity



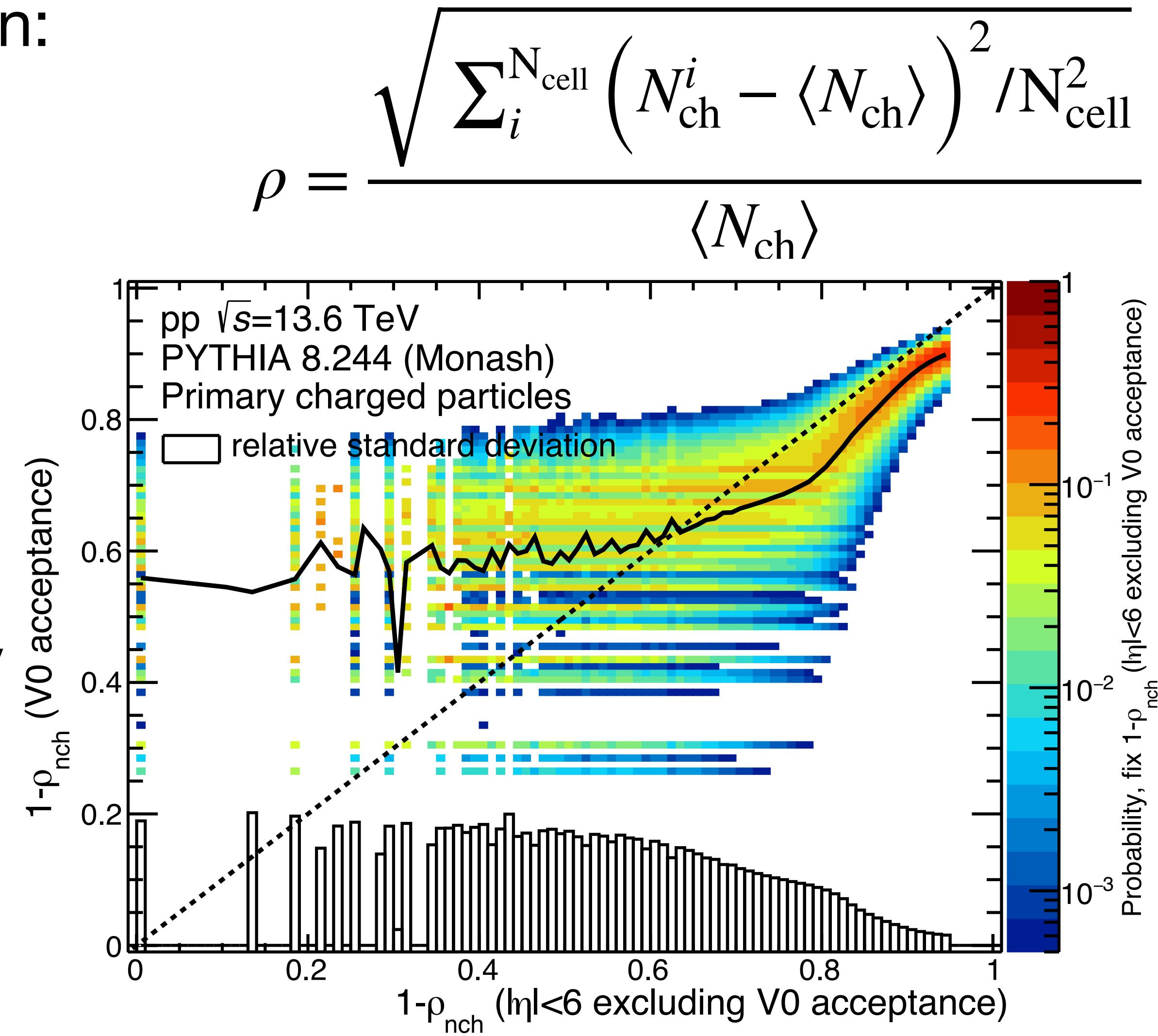
- Flattenicity definition:
- Based on MC simulations, flattenicity in the pseudorapidity interval covered by ALICE V0A and V0C detectors is strongly correlated with the global shape of the event



# Flattenicity



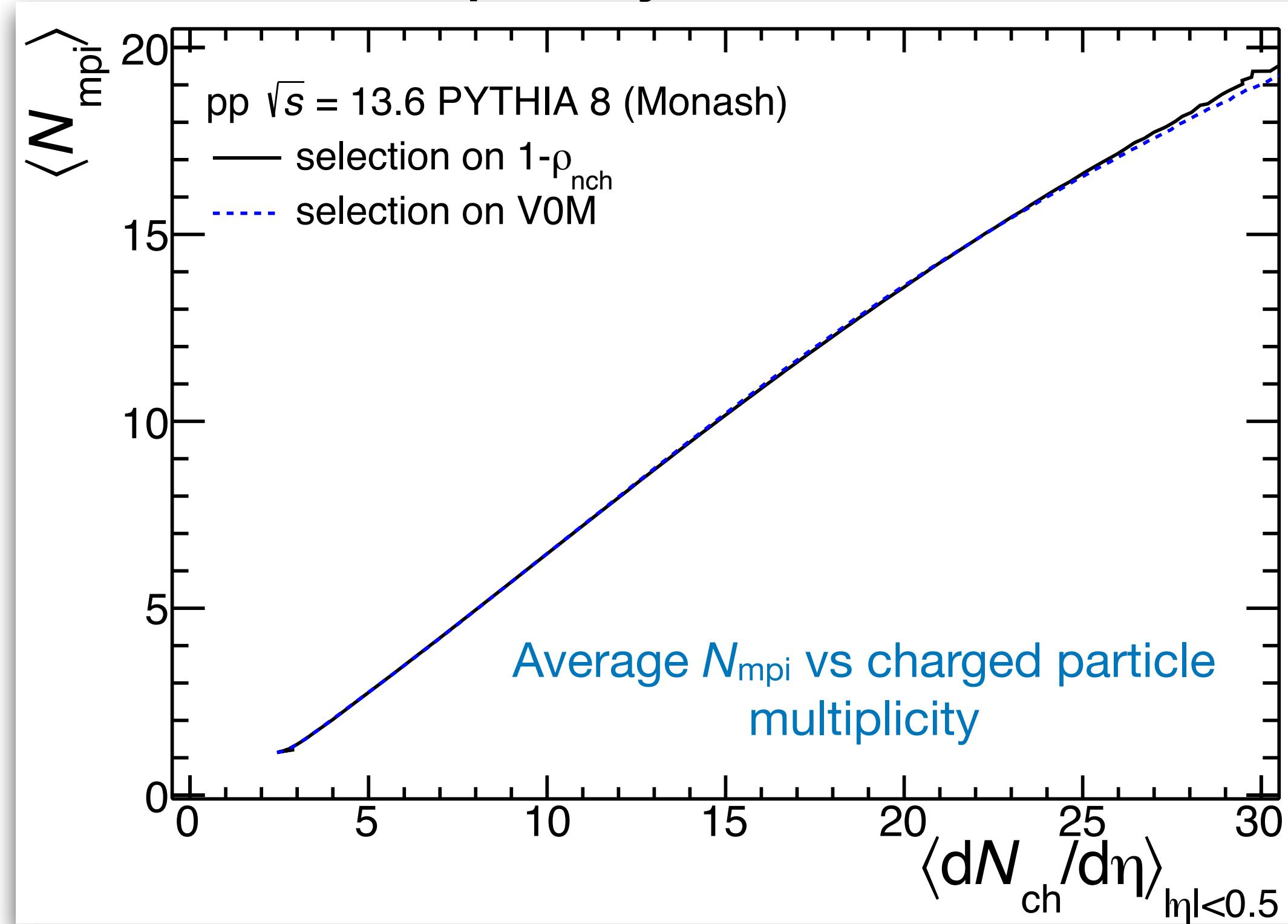
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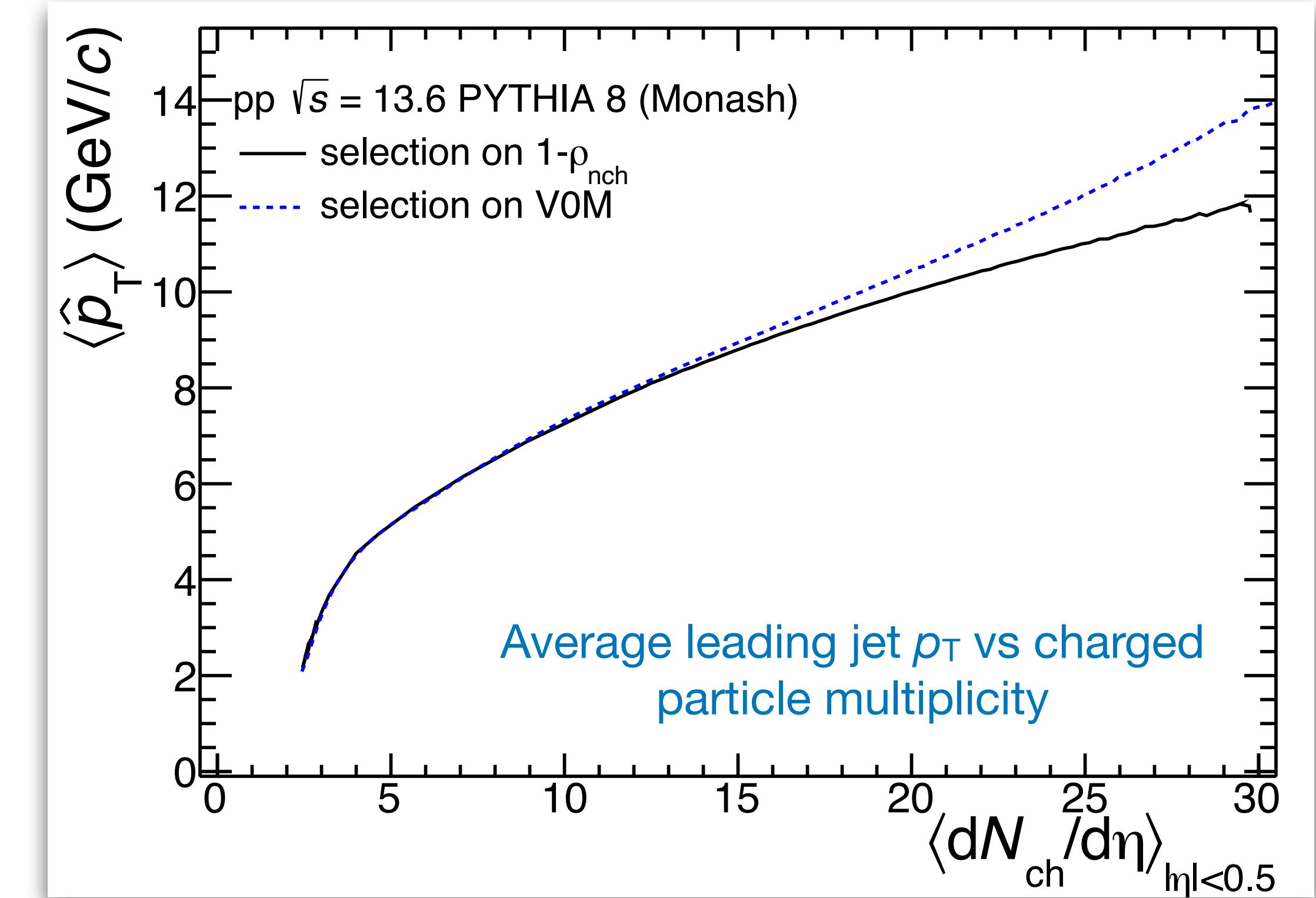
Andreas' request: check the correlation excluding the V0 acceptance

# Sensitivity of flattenicity to MPI

- Same sensitivity to MPI as the V0M multiplicity estimator

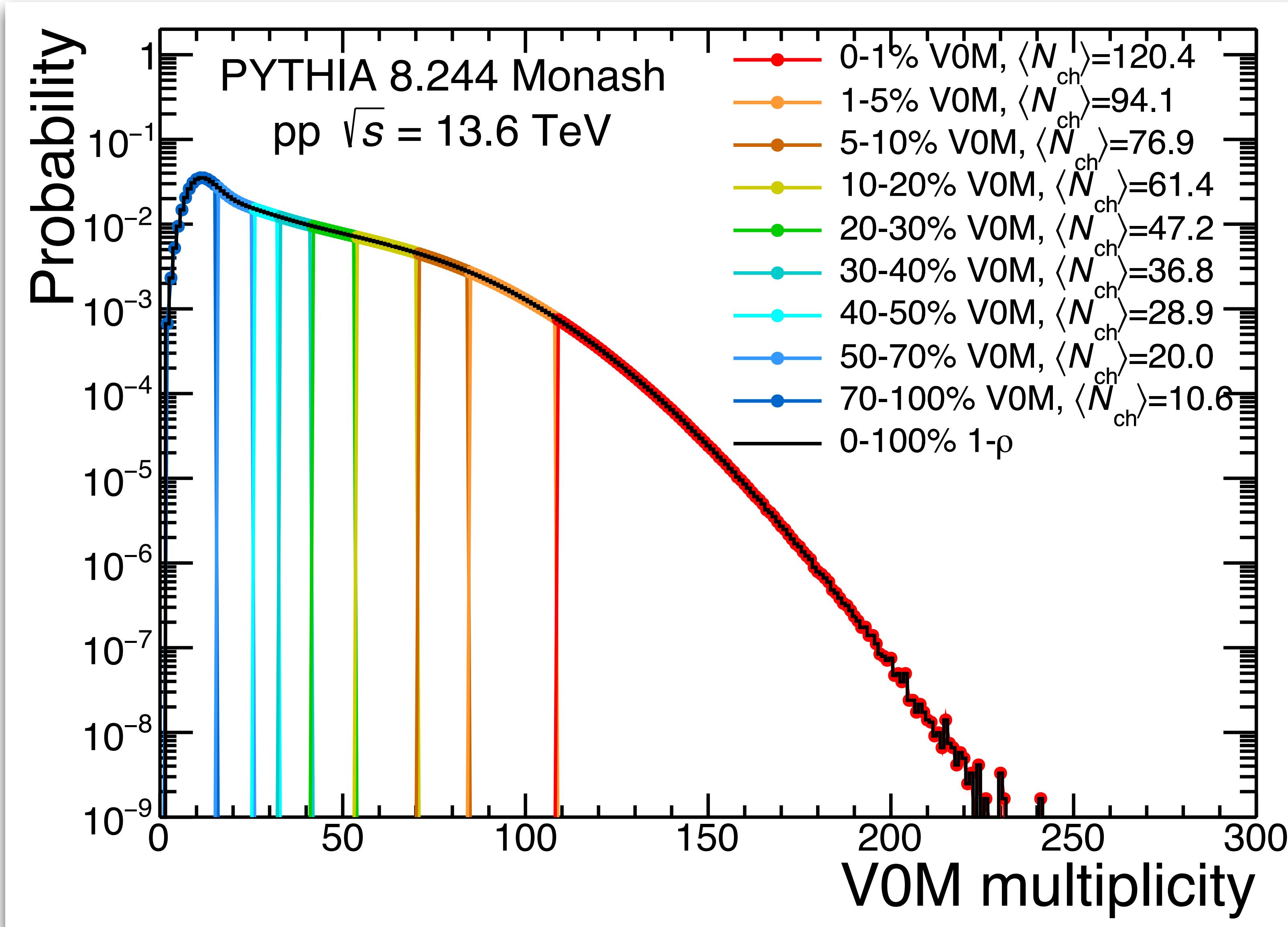


- Flattenicity selects “softer” pp collisions than the V0M estimator

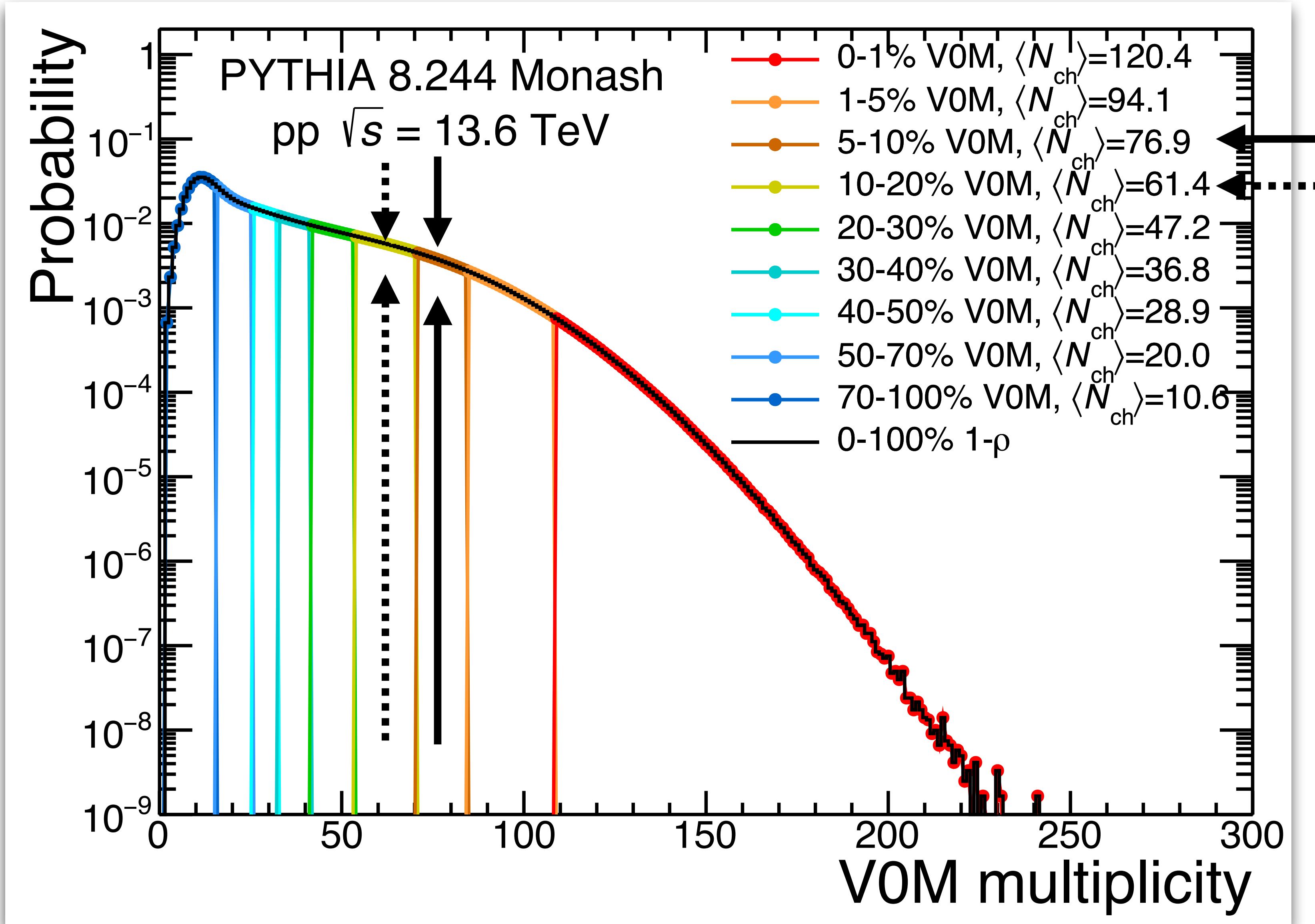


See A. Ortiz, A. Khuntia, O. Vázquez, S. Tripathy, G. Bencédi, S. Prasad and F. Fan, “Unveiling the effects of multiple soft partonic interactions in pp collisions at 13.6 TeV using a new event classifier” to appear in Phys. Rev. D [2211.06093]

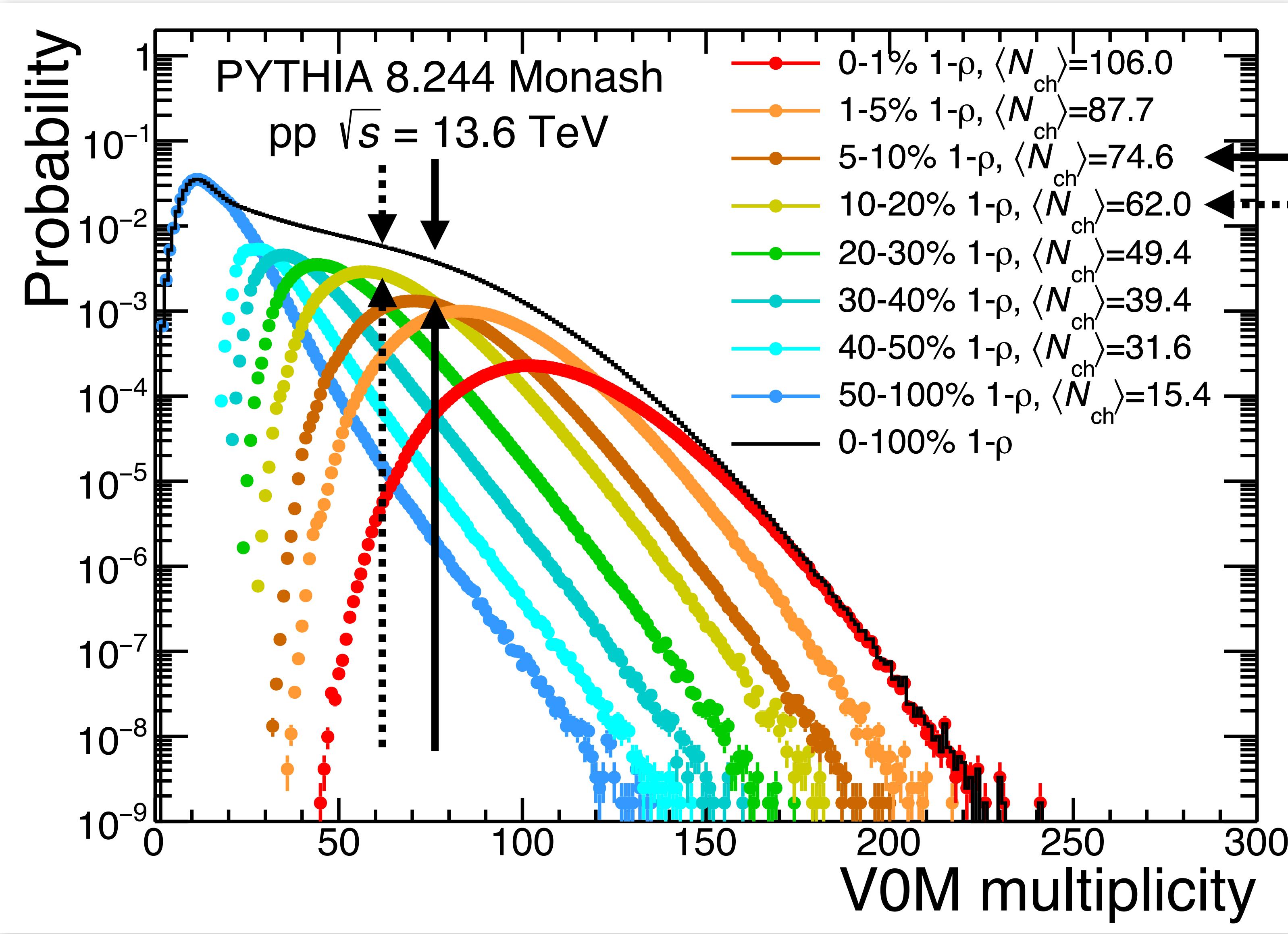
# VOM event classes



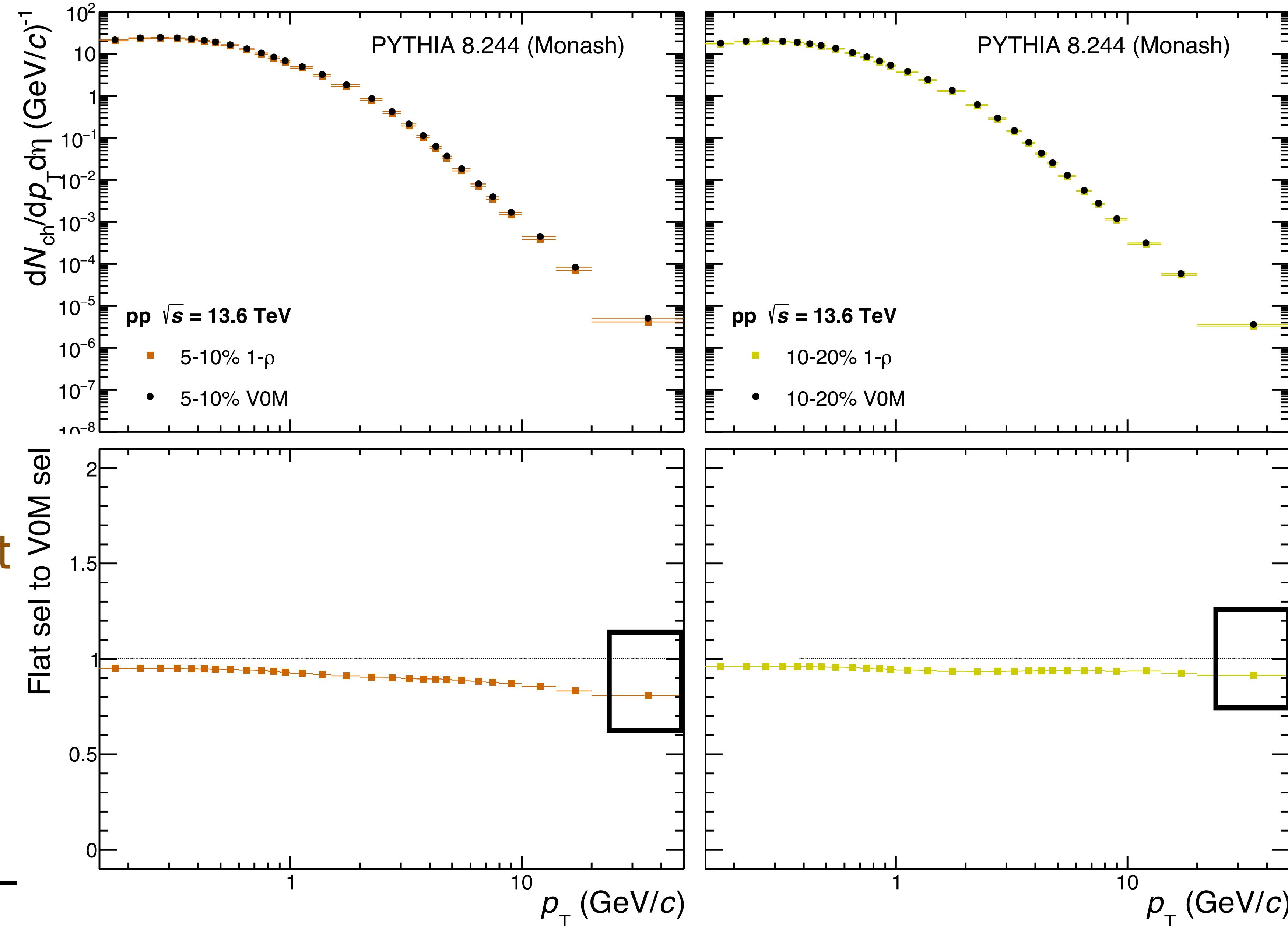
# VOM event classes



# Flattenicity event classes



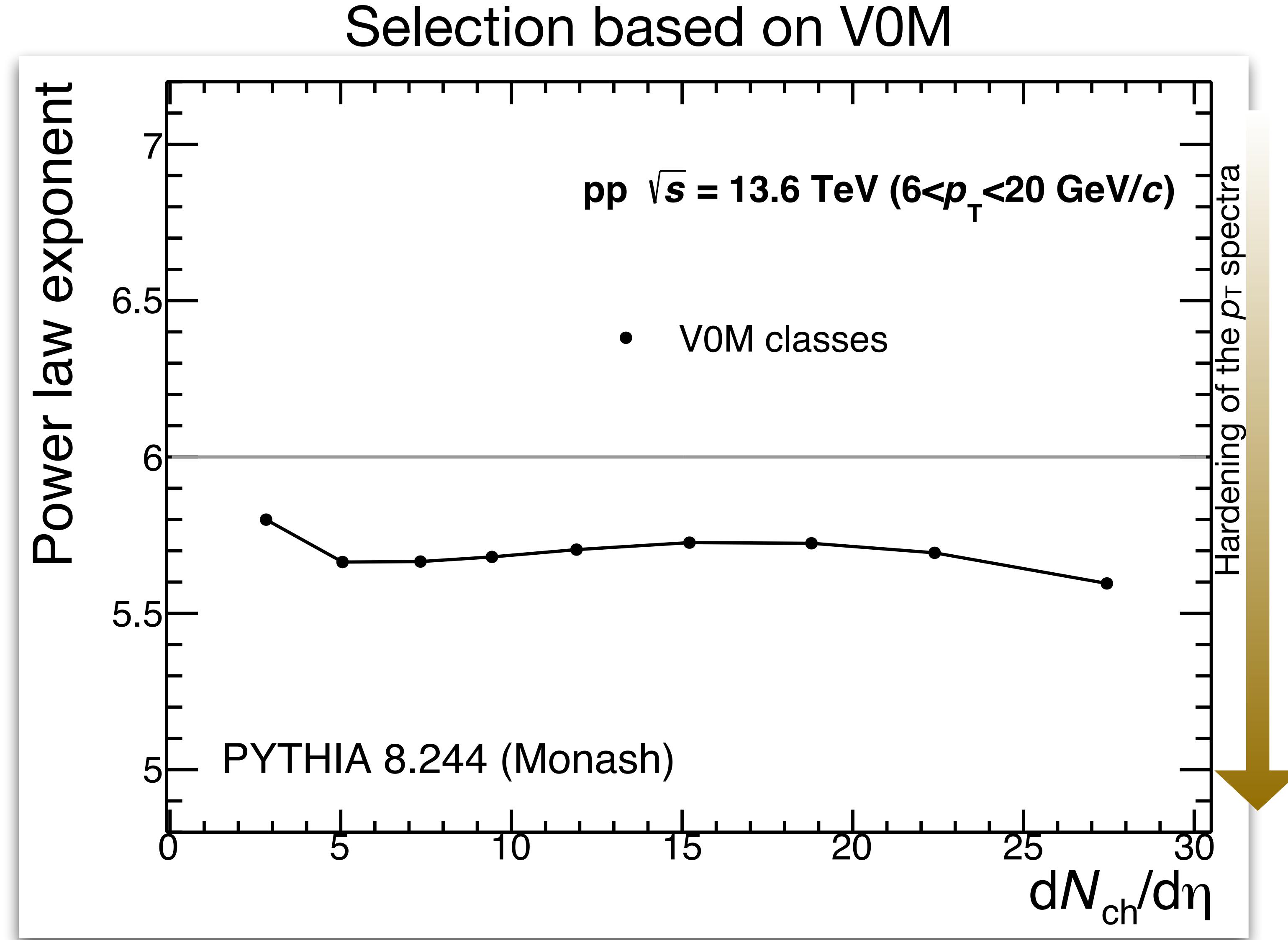
# Event classes with similar $\langle N_{ch} \rangle$



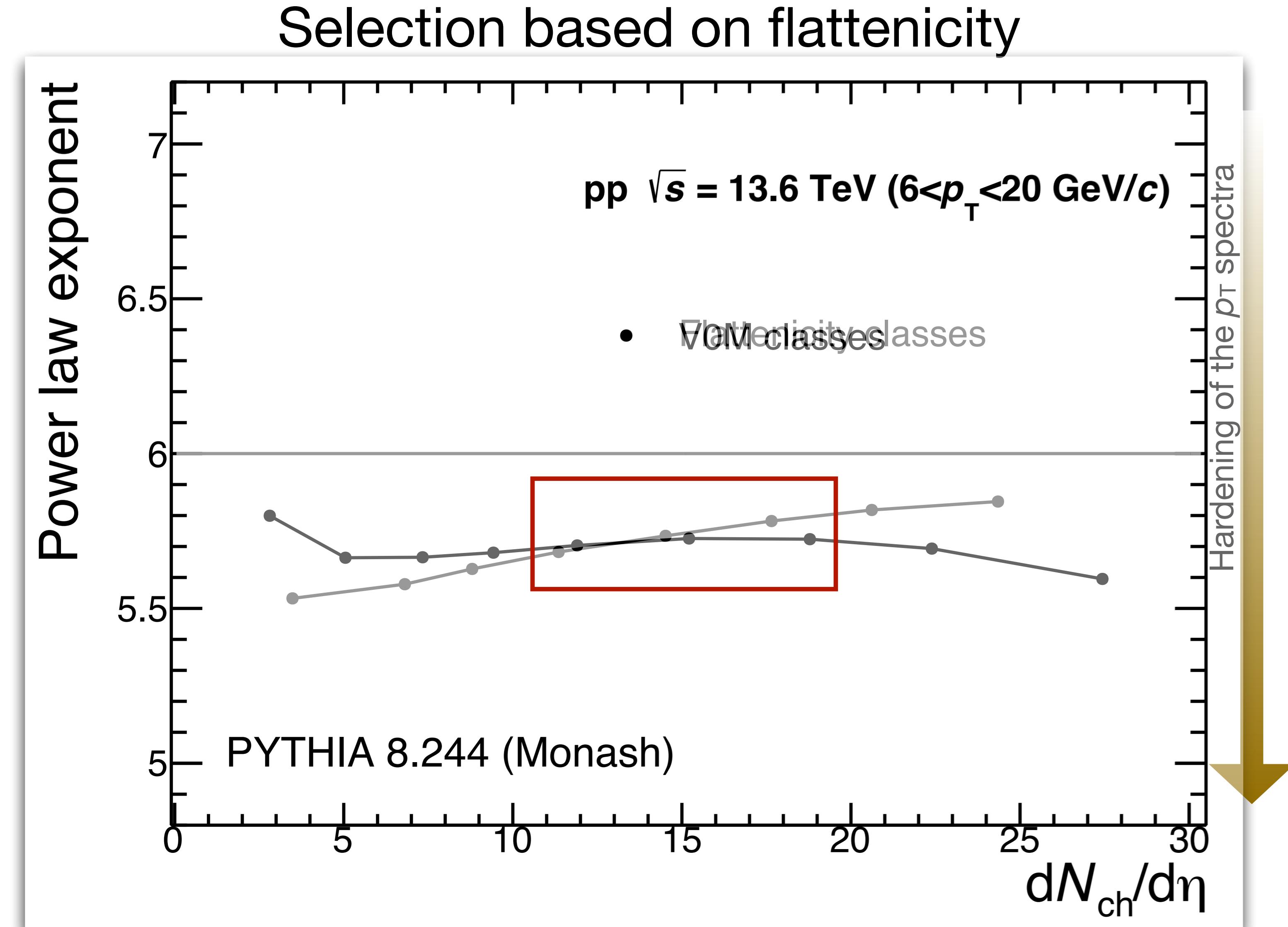
Diference in  
V0M mult:  
~3%  
Difference at  
high  $p_T$ :  
>15%

Diference in  
V0M mult:  
~1%  
Difference at  
high  $p_T$ :  
>5%

# High- $p_T$ yield



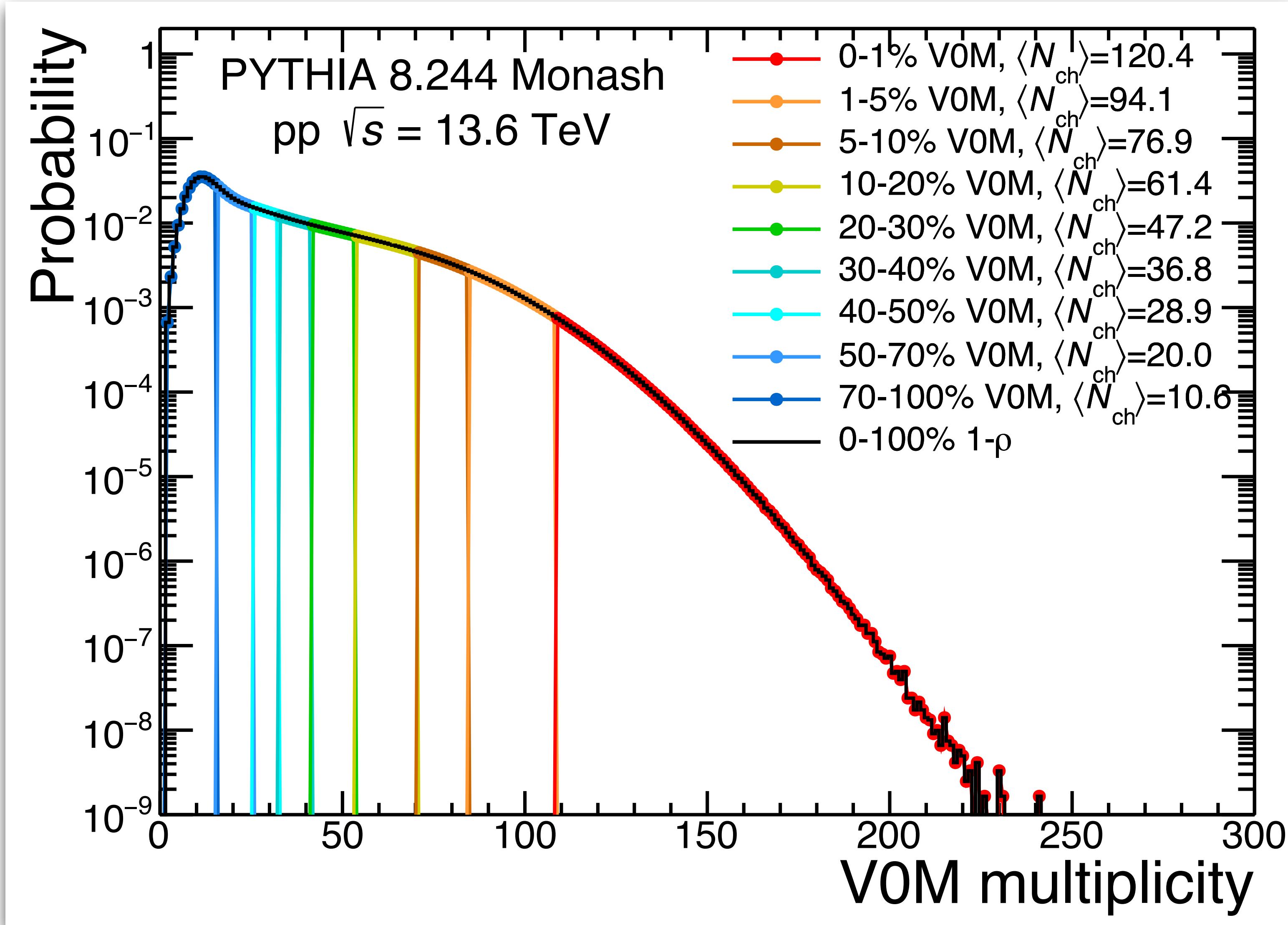
# High- $p_T$ yield



Softer interactions are selected using flattenicity than with V0M multiplicity

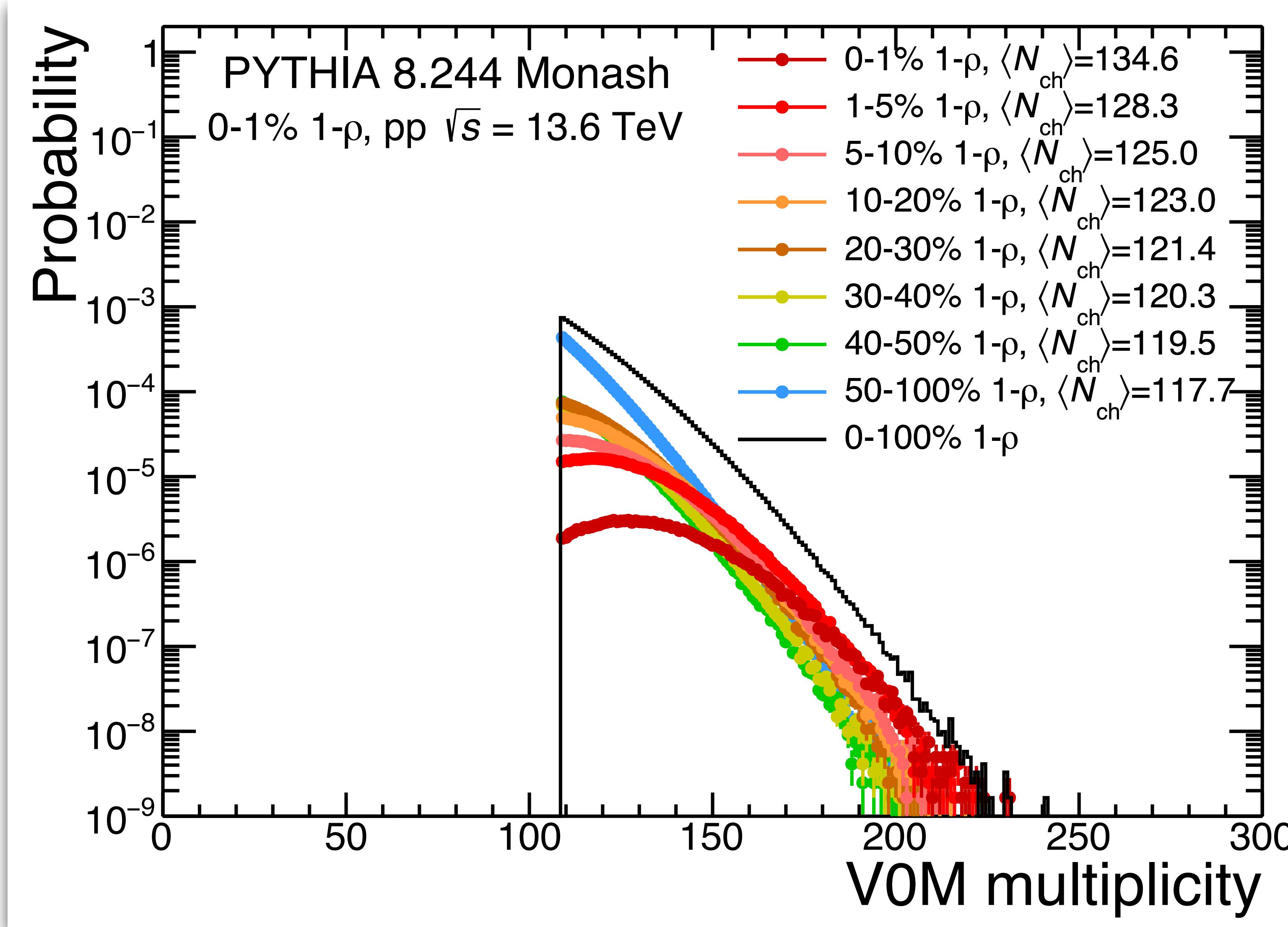
# Double differential analysis

## Selection based on VOM



# Double differential analysis

## Selection based on VOM + flattenicity (example 0-1% VOM)

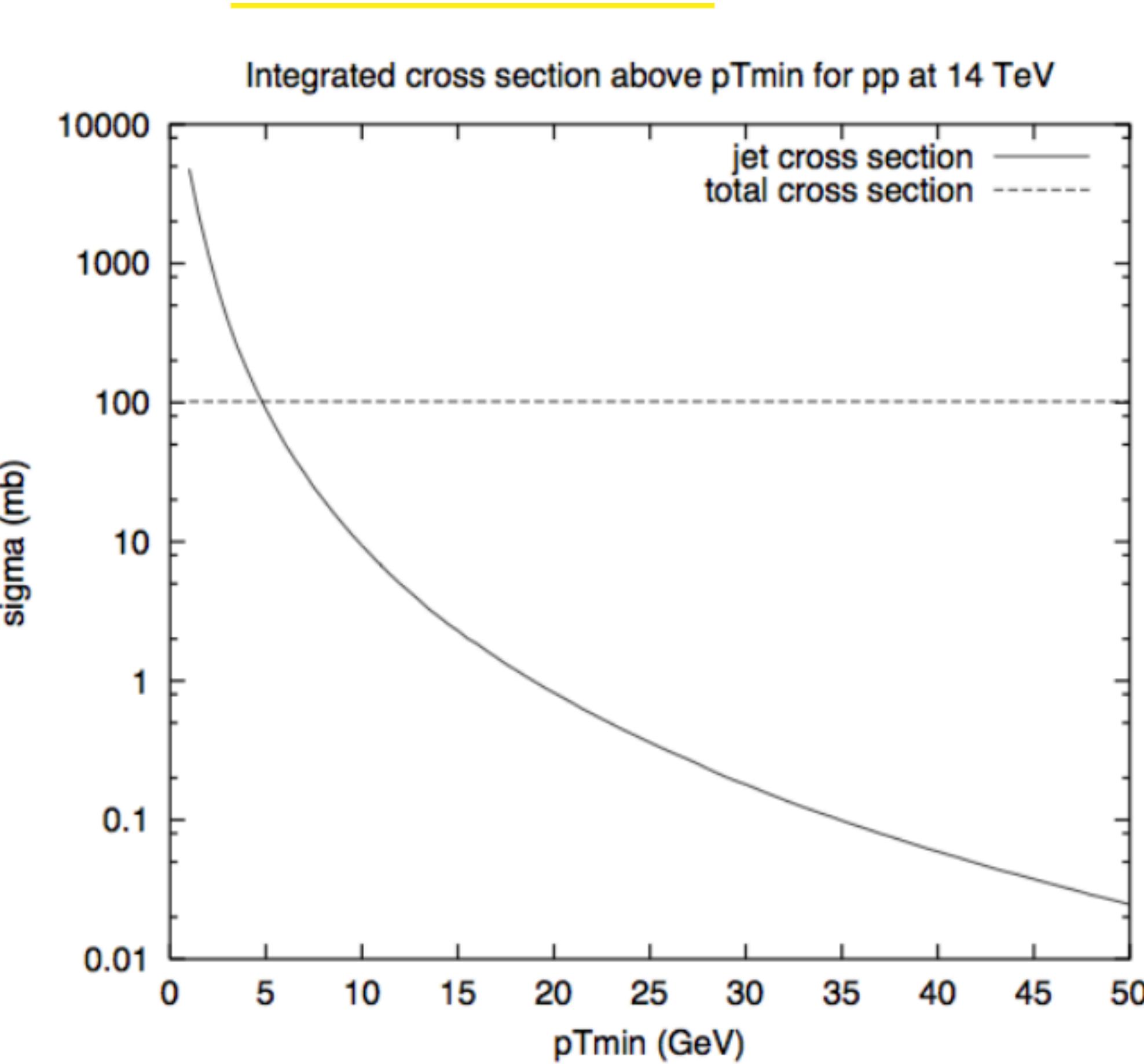
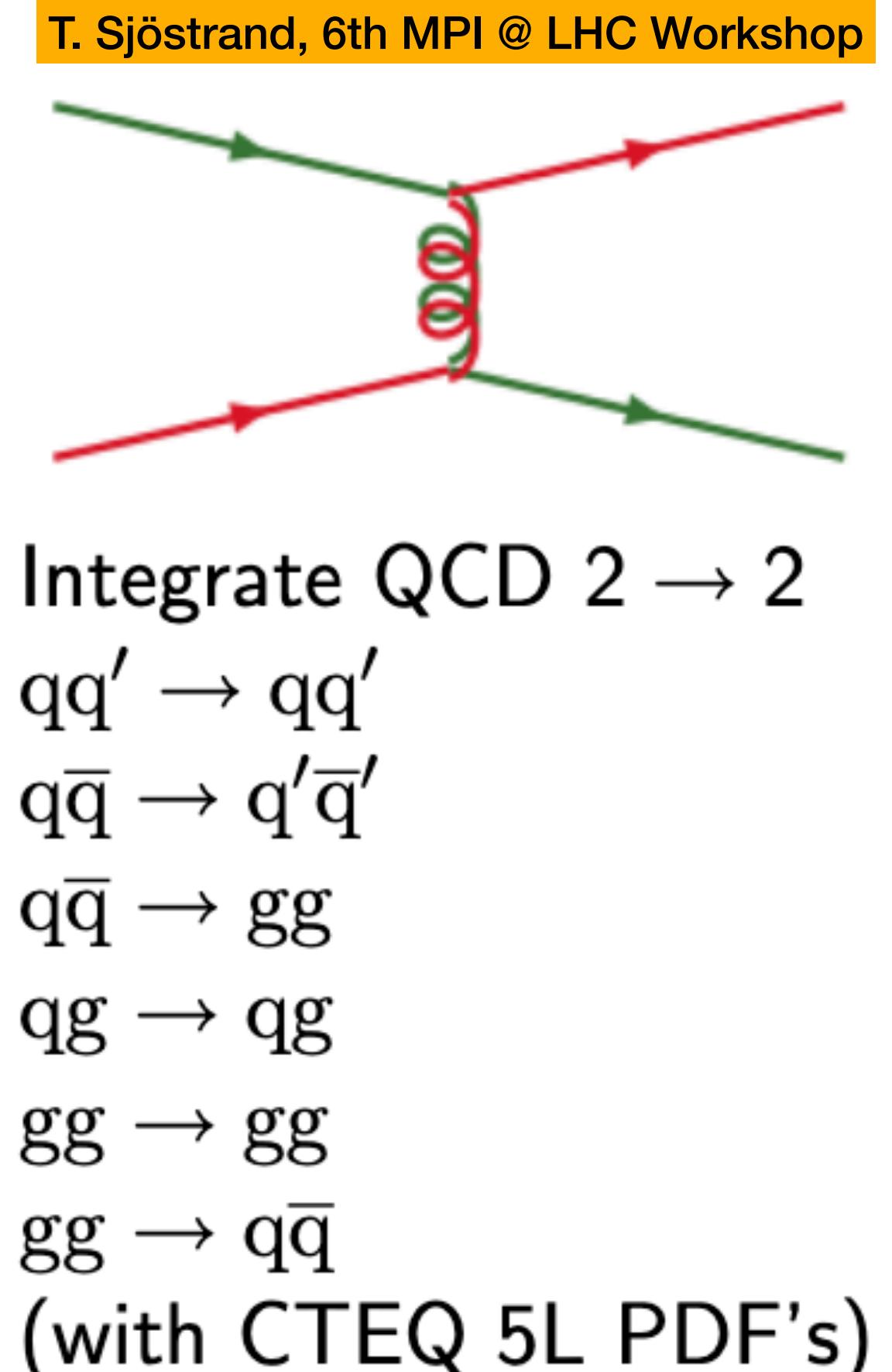


High flattenicity events (0-1% 1- $\rho$ ) do not correspond to the highest multiplicity class. Instead, flattenicity exhibit a modest multiplicity dependence

# Multiparton interactions MPI

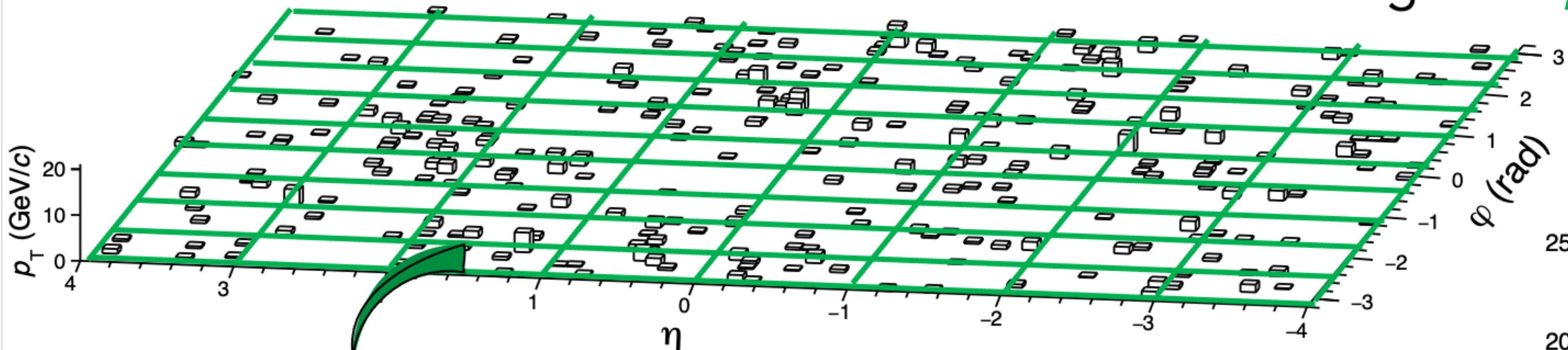
At high energies, the leading order cross-section for  $2 \rightarrow 2$  parton scatterings with momentum transfer

$Q > Q_{\min} \gg \Lambda_{\text{QCD}}$  exceeds the total pp cross-section at a range of  $Q_{\min}$ -values where perturbative QCD is applicable (at LHC,  $Q_{\min} \approx 4$  GeV/c) [T. Sjöstrand and M. Zijl Phys. Rev. D36 (1987)]



# Flattenicity

PYTHIA 8.303 (Monash 2013), pp  $\sqrt{s} = 13$  TeV,  $N_{\text{mpi}}=24$ ,  $N_{\text{ch}}=325$

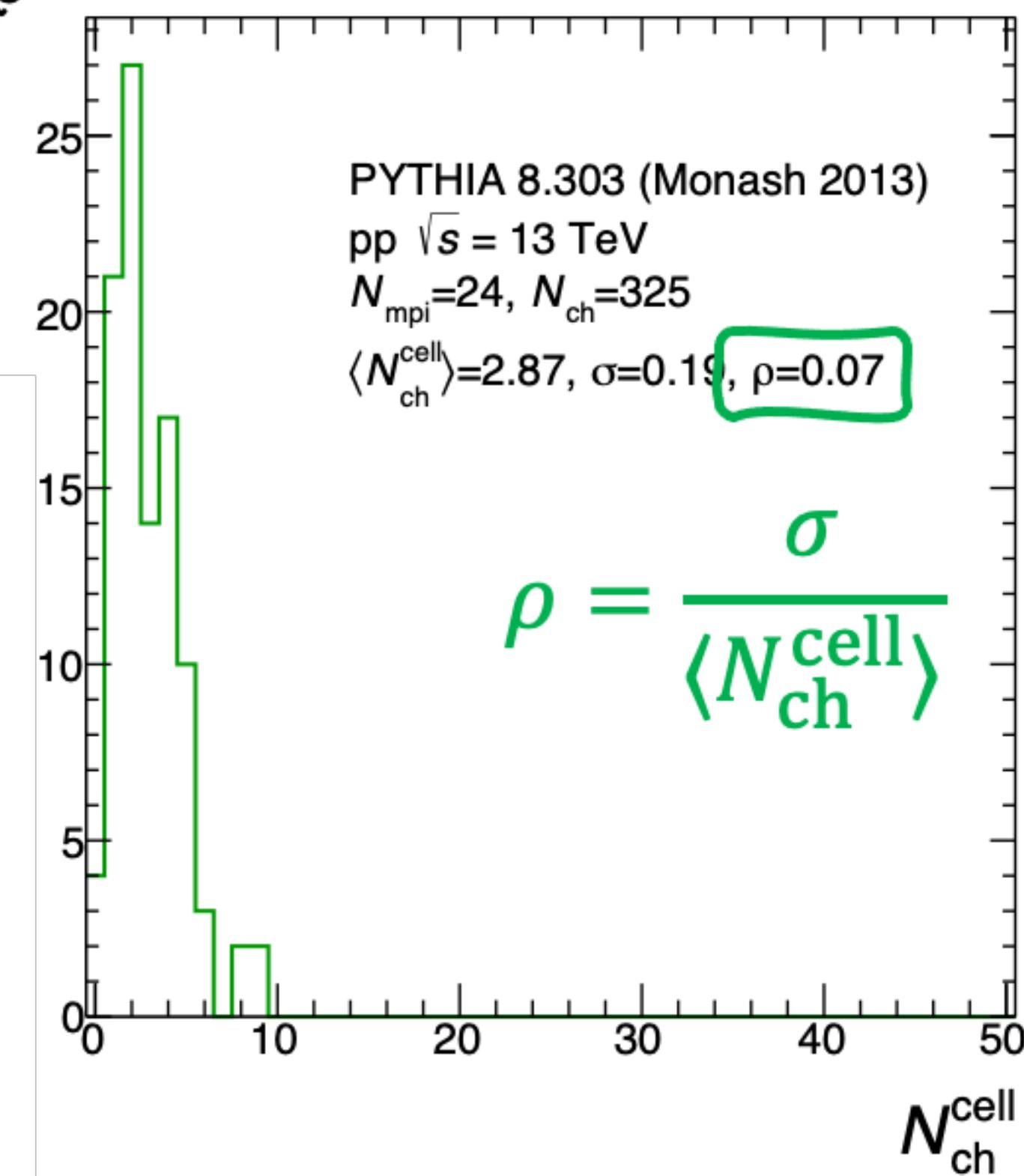


A grid in  $\varphi - \eta$  is built:  $10 \times 8$

In each cell, the charged particle multiplicity is computed:  $N_{\text{ch}}^{\text{cell}}$

EBE the  $N_{\text{ch}}^{\text{cell}}$  distribution is obtained

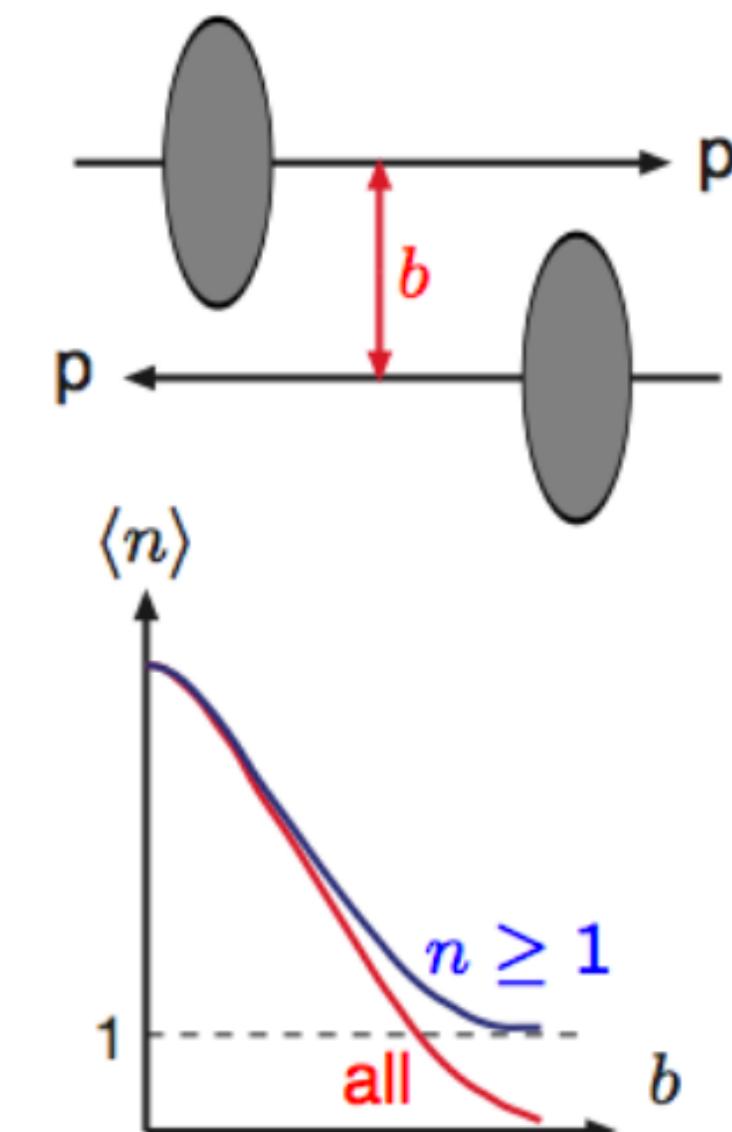
Events with several minijets (very active MPI) are expected to have small  $\rho$  values



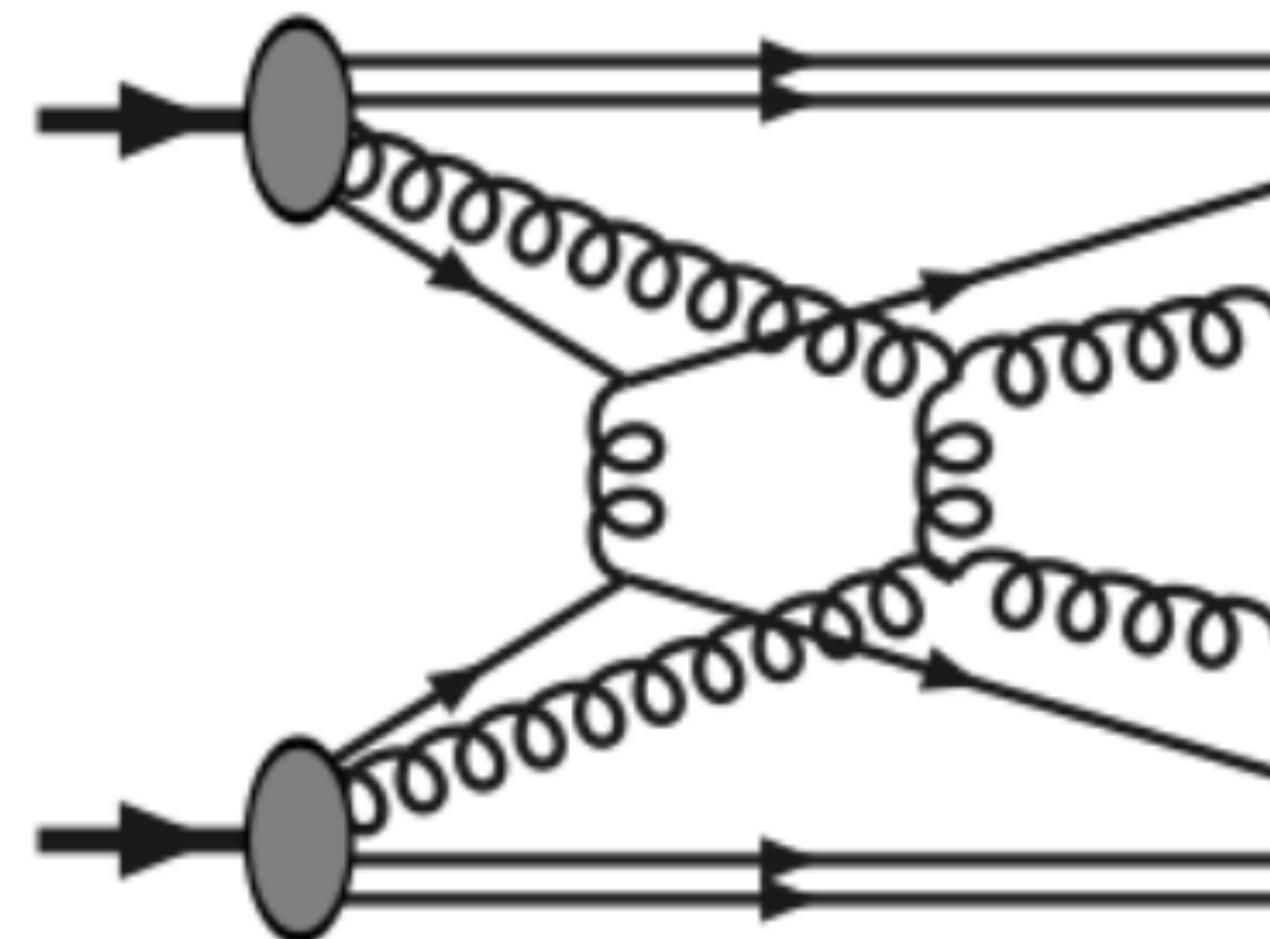
At high energy, the cross section for a 2 → 2 parton interaction with momenta  $p$  and  $Q > Q_{\min}$  is the total pp cross section over the range of  $Q$  where perturbative calculations are applicable ( $\sim 10$  GeV/c) [T. Sjöstrand, Zijl Phys. Rev. D 55, 303 (1997)]

## Interpretation: Many partonic scatterings per event: (MPI)

- MPI is a logical consequence of the composite nature of protons



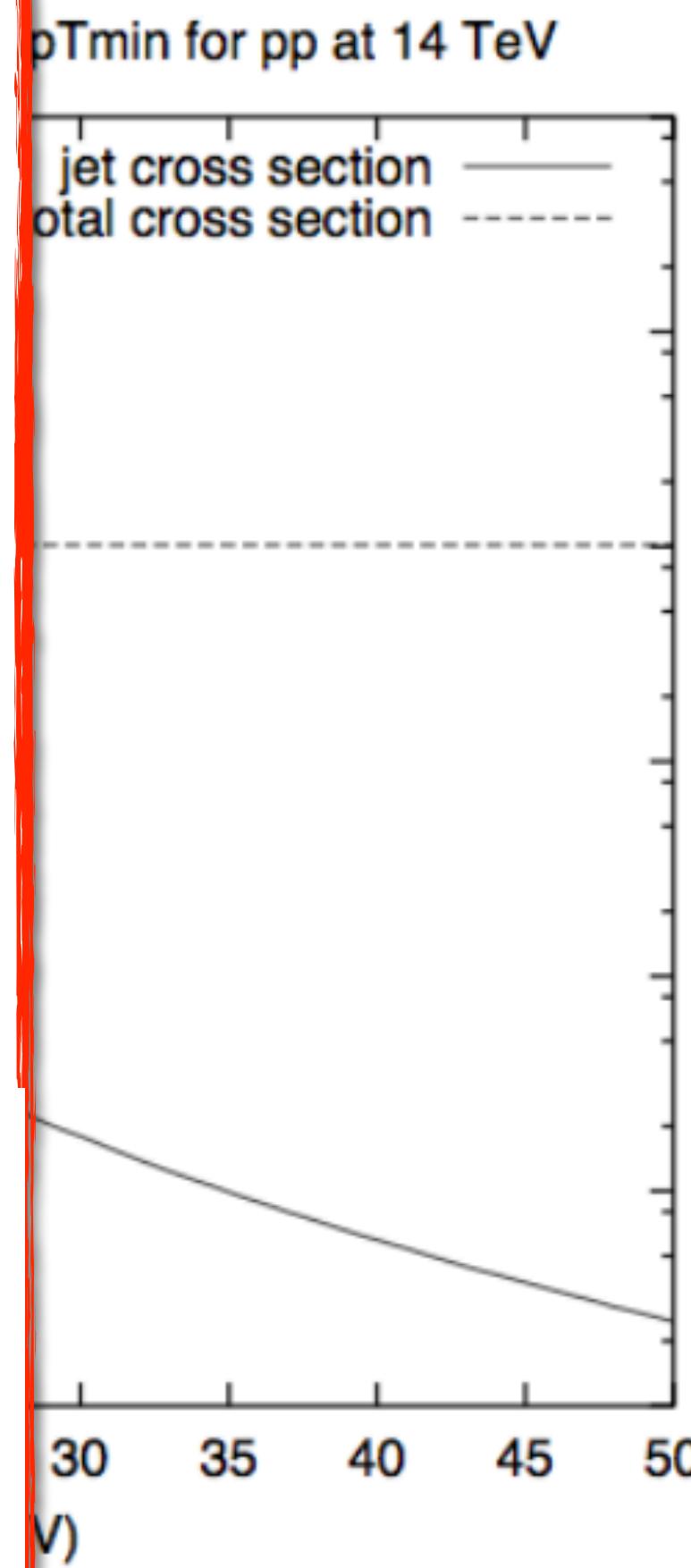
- In event generators like Pythia, an impact parameter dependence is considered



Overlap of protons during encounter is

$$\mathcal{O}(b) = \int d^3x dt \rho_1(x, t) \rho_2(x, t)$$

where  $\rho$  is (boosted) matter distribution in  $p$ , e.g. Gaussian or more narrow peak.



T. Sjöstrand, ISAPP 2018