

## Multiparton Interactions in pp collisions from Machine Learning

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**Erik Zepeda\*** and **Antonio Ortiz**

*Instituto de Ciencias Nucleares, Universidad Nacional Autónoma de México,  
Apartado Postal 70-543, México Distrito Federal 04510, México*

*E-mail: [eazg@ciencias.unam.mx](mailto:eazg@ciencias.unam.mx), [antonio.ortiz@nucleares.unam.mx](mailto:antonio.ortiz@nucleares.unam.mx)*

Over the last years, Machine Learning (ML) tools have been successfully applied to a wealth of problems in high-energy physics. In this work, we discuss the extraction of the average number of Multiparton Interactions ( $\langle N_{\text{mpi}} \rangle$ ) from minimum-bias pp data at LHC energies using ML methods. Using the available ALICE data on transverse momentum spectra as a function of multiplicity, we report that for minimum-bias pp collisions at  $\sqrt{s} = 7$  TeV the average  $N_{\text{mpi}}$  is  $3.98 \pm 1.01$ , which complements our previous results for pp collisions at  $\sqrt{s} = 5.02$  and 13 TeV. The comparisons indicate a modest energy dependence of  $\langle N_{\text{mpi}} \rangle$ . We also report the multiplicity dependence of  $N_{\text{mpi}}$  for the three center-of-mass energies. These results are qualitatively consistent with the existing ALICE measurements sensitives to MPI, therefore they provide additional experimental evidence of the presence of MPI in pp collisions.

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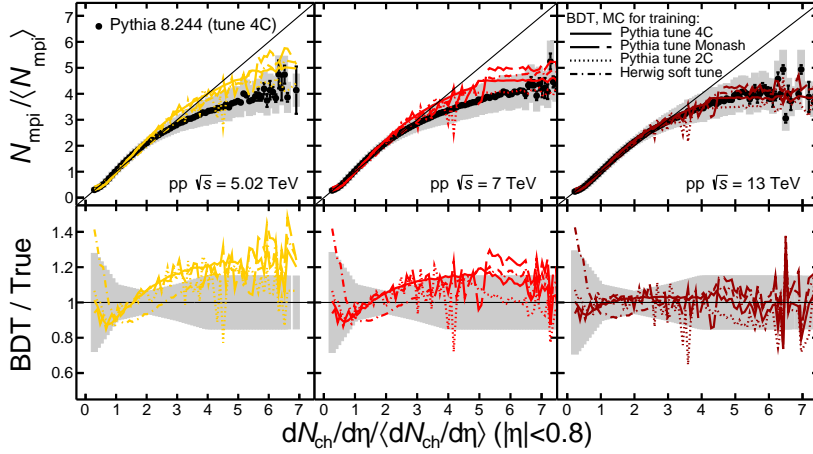
\*Speaker

## 1. Introduction

The study of Multiparton Interactions (MPI) in pp collisions has recently attracted the attention of the heavy-ion community, because surprisingly, the high-multiplicity pp data unveiled heavy-ion-like features, i.e. azimuthal anisotropies [1], the enhancement of (multi-)strange hadrons [2], as well as radial flow patterns in the transverse momentum ( $p_T$ ) spectra of identified hadrons [3]. Besides the hydrodynamical approach [4, 5], MPI, offers an alternative possibility to explain the observed phenomena. For instance, color reconnection and MPI can mimic radial flow patterns in pp collisions [6]. In this direction, we have proposed the extraction of the MPI activity from minimum-bias pp data using Machine Learning (ML) methods [7, 8]. In this contribution, we summarize the main results including the multiplicity dependence of the average number of MPI extracted from the available ALICE data at the LHC [3, 9].

## 2. Analysis

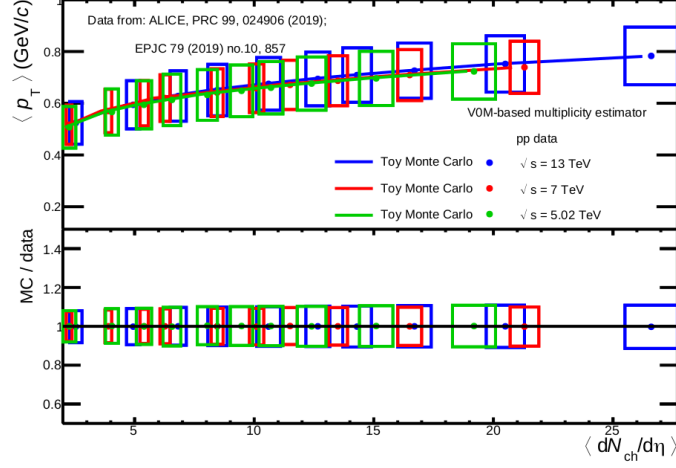
Our approach relies on a multivariate regression technique based on Boosted Decision Trees (BDT). The study is conducted using the Toolkit for Multivariate Analysis (TMVA) framework which provides a ROOT-integrated ML environment for the processing and parallel evaluation of multivariate classification and regression techniques [10]. The training is performed using pp collisions at  $\sqrt{s} = 13$  TeV simulated with PYTHIA 8.244 [11] tune 4C [12]. The choice of the input variables is based on their correlation with  $N_{\text{mpi}}$  [13]. We consider the event-by-event average transverse momentum and the mid-pseudorapidity charged particle multiplicity ( $N_{\text{ch}}$ ).



**Figure 1:** Monte Carlo closure test using pp collisions at  $\sqrt{s} = 5.02$  (left), 7 (middle) and 13 TeV (right) simulated with PYTHIA 8 tune 4C. The top panels display the self normalized average number of MPI as a function of the self-normalized mid-pseudorapidity charged particle multiplicity. Ratios between ML results and the true values given by PYTHIA are shown in the bottom panels.

The systematic uncertainties take into account a variation of the PYTHIA 8 tune, as well as the MPI and hadronization model. To this end, tunes: 2C, 4C and Monash 2013 were used for training, and the effects of MPI and hadronization was investigated using the Monte Carlo (MC) generator HERWIG 7.2 [14] for training. Figure 1 shows the correlation between the self normalized number

of MPI ( $N_{\text{mpi}}/\langle N_{\text{mpi}} \rangle$ ) and the self normalized mid-pseudorapidity charged particle multiplicity ( $N_{\text{ch}}/\langle N_{\text{ch}} \rangle$ ) in pp collisions at  $\sqrt{s} = 5.02, 7$  and  $13$  TeV. For  $N_{\text{ch}}/\langle N_{\text{ch}} \rangle < 3$ , the self normalized  $N_{\text{mpi}}$  increases linearly with the event multiplicity. On the other hand, for higher multiplicities, we observe a deviation of the self normalized  $N_{\text{mpi}}$  with respect to the linear trend. The Figure 1 also displays the results obtained from regression (lines), and shows that using ML-based regression, one can recover the energy and multiplicity dependence.

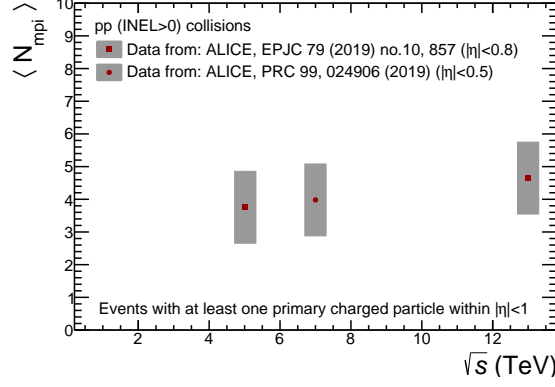


**Figure 2:** Mean transverse momentum as a function of the average charged-particle multiplicity density in pp collisions at  $\sqrt{s} = 5.02, 7$  and  $13$  TeV. In the top panel ALICE data [3, 9] (solid markers) are compared with results from the Toy MC (solid lines). Bottom panel displays ratios between Toy MC and the data.

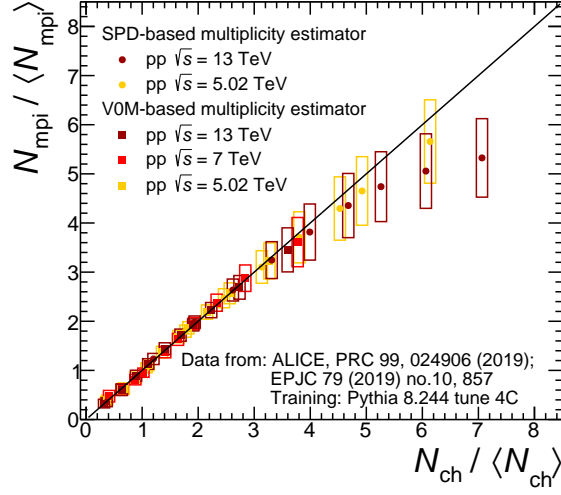
Regarding the analysis using available data, we built a toy MC in order to get the correlation between the event-by-event  $\langle p_T \rangle$  and  $N_{\text{ch}}$ . Figure 2 displays the mean transverse momentum as a function of the average charged-particle multiplicity density in pp collisions at  $\sqrt{s} = 5.02, 7$  and  $13$  TeV. Within uncertainties, the toy MC reproduces the correlation between the  $\langle p_T \rangle$  and  $\langle dN_{\text{ch}}/d\eta \rangle$ . In our approach, the event-by-event information produced by the toy MC was processed with the trained BDT in order to extract the MPI activity associated with the data.

### 3. Results

Using the ALICE data from pp collisions at  $\sqrt{s} = 7$  TeV [3], we extract the average number of MPI, which is found to be  $\langle N_{\text{mpi}} \rangle = 3.98 \pm 1.01$ . Figure 3 displays the average number of MPI as a function of the center-of-mass energy, for pp collision at  $\sqrt{s} = 5.02, 7$  and  $13$  TeV. We obtain a regression value which is above unity, therefore, our results support the presence of MPI in pp collisions. We also observe a modest energy dependence, which is similar to that predicted by PYTHIA 8 [7]. A similar finding has been discussed in Ref. [15], where the energy dependence of underlying-event observables has been reported. Figure 4 displays the self normalized number of MPI ( $N_{\text{mpi}}/\langle N_{\text{mpi}} \rangle$ ) as a function of the self-normalized mid-pseudorapidity charged-particle multiplicity ( $N_{\text{ch}}/\langle N_{\text{ch}} \rangle$ ) in pp collisions at  $\sqrt{s} = 5.02, 7$  and  $13$  TeV from ALICE data. We observe that for  $N_{\text{ch}} < 3\langle N_{\text{ch}} \rangle$  the self normalized  $N_{\text{mpi}}$  increases linearly with the event multiplicity. Regarding higher multiplicities, we observe a deviation of the self normalized  $N_{\text{mpi}}$  with respect to the linear trend. This result qualitatively agrees with PYTHIA 8 (see figure 1).



**Figure 3:** Average number of MPI as a function of the center-of-mass energy. Results for pp collisions at  $\sqrt{s} = 7$  TeV, are compared to those for pp collisions at  $\sqrt{s} = 5.02$  and 13 TeV reported in [7].



**Figure 4:** The self normalized average number of Multiparton Interactions as a function of the self normalized mid-pseudorapidity charged particle multiplicity is shown for pp collisions at  $\sqrt{s} = 5.02$ , 7 and 13 TeV.

#### 4. Conclusions

We report the extraction of the average number of MPI from pp data at the LHC energies. We have found  $\langle N_{\text{mpi}} \rangle = 3.98 \pm 1.01$  for pp collisions at  $\sqrt{s} = 7$  TeV. The comparisons with our previous results for pp collisions at  $\sqrt{s} = 5.02$  and 13 TeV indicate a modest energy dependence of  $N_{\text{mpi}}$ . This result provide experimental evidence of the presence of MPI in hadronic interactions. In addition, we also report the multiplicity dependence of  $N_{\text{mpi}}$  for the three center-of-mass energies. Our results are fully consistent with the so-called “mini-jet analysis” of ALICE [16], and suggest that high multiplicities (at mid-pseudorapidity) can only be reached by selecting events with many high-multiplicity jets.

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