Inference of the Mass Composition of Cosmic Rays with energies from 10^{18.5} to 10²⁰ eV using the Pierre Auger Observatory and Deep Learning

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We present measurements of the atmospheric depth of the shower maximum X_{max} , inferred for the first time on an event-by-event level using the Surface Detector of the Pierre Auger Observatory. Using deep learning, we were able to extend measurements of the X_{max} distributions up to energies of 100 EeV (10^{20} eV), not yet revealed by current measurements, providing new insights into the mass composition of cosmic rays at extreme energies. Gaining a 10-fold increase in statistics compared to the Fluorescence Detector data, we find evidence that the rate of change of the average X_{max} with the logarithm of energy features three breaks at 6.5 ± 0.6 (stat) ± 1 (sys) EeV, 11 ± 2 (stat) ± 1 (sys) EeV, and 31 ± 5 (stat) ± 3 (sys) EeV, in the vicinity to the three prominent features (ankle, instep, suppression) of the cosmic-ray flux. The energy evolution of the mean and standard deviation of the measured X_{max} distributions indicates that the mass composition becomess increasingly heavier and purer, thus being incompatible with a large fraction of light nuclei between 50 EeV and 100 EeV.

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INTRODUCTION

The arrival directions, energy spectrum, and mass composition are the three important pillars of cosmic ray research. A sound interpretation of the three measurements and their energy dependence, both individually and jointly, is pivotal for a deep understanding of the nature of cosmic rays, including their origin and propagation, and enables the study of astrophysical models. With energies larger than 1 EeV (10^{18} eV), ultra-high-energy cosmic rays (UHECRs) are the most energetic particles ever measured by humankind. One of the lasting puzzles is the origin of the suppression of the cosmic-ray flux observed at around 50 EeV[1-4]. A precise measurement of the UHECR mass composition can deliver insights into whether the suppression is caused by the interaction of the particles with the cosmic background photons [5, 6], a sign of the maximum energy reached in cosmic accelerators [7], or a combination of both [8, 9]. Due to the low flux at ultrahigh energies, the primary composition cannot be measured directly but can only be studied by indirectly analyzing the properties of the induced air showers. Information on the primary mass can be obtained by measuring the atmospheric depth of the shower maximum X_{max} , the depth at which the number of secondary particles reaches its maximum. Investigating the measured X_{max} distribution, as a function of energy, in terms of its mean and standard deviation (fluctuations), $\langle X_{\text{max}} \rangle$ and $\sigma(X_{\text{max}})$, enables us to study the UHECR mass composition [10]. Heavier particles feature, on average, a smaller X_{max} since sub-showers are sharing the primary energy, resulting in a maximum higher in the atmosphere. This motivates the investigation of the first moment $\langle X_{\text{max}} \rangle$ of the X_{max} distribution. Further, the shower-to-shower fluctuations, i.e., the second moment $\sigma(X_{\text{max}})$ of the distribution, is also a mass-sensitive observable. Due to the smaller cross-section and the development of fewer sub-showers, cascades induced by lighter primary particles are subject to larger fluctuations. Furthermore, compared to $\langle X_{\text{max}} \rangle$, the fluctuations $\sigma(X_{\text{max}})$ are sensitive to both the primary mass and the degree of mixing of the primary beam [11] and are almost insensitive to the uncertainties in the hadronic interaction models.

Using fluorescence telescopes, X_{max} can be directly reconstructed by observing the longitudinal shower development. Nevertheless, due to the observations being confined to dark and moonless nights, the duty cycle is limited. In contrast, sparse surface-detector arrays have a duty cycle close to 100% and sample the secondary shower particles at the ground. Thus, they cannot directly observe X_{max} , making its reconstruction challenging. However, information about the shower development and X_{max} is contained in the lateral number density and distribution of arrival times of particles reaching the ground. By studying the risetimes of the time-dependent signals, conclusions on the average composition have already been drawn in the past [12]. However, to infer the UHECR mass composition beyond mere $\langle X_{\text{max}} \rangle$ measurements, more sophisticated methods are needed to fully exploit the complex data. The advent of deep learning [13, 14] provides new analysis techniques for large and complex data sets. First approaches have already been successfully applied to LHC data [15] and physics in general [16]. The recent progress offers supplementary and improved reconstruction algorithms for neutrino [17–19] and cosmic-ray observatories [20]. This includes the deep-learning-based reconstruction of X_{max} [21–24] and muon signals [25] using the temporal structure of signals measured by the Surface Detector of the Pierre Auger Observatory.

In this work, we use this novel reconstruction technique to study the mass composition of UHECRs in terms of $\langle X_{\text{max}} \rangle$ and $\sigma(X_{\text{max}})$ in the energy range from 3 to 100 EeV. With about 50,000 events, this is, at the time, the most comprehensive study of the UHECR mass composition and the first measurement of $\sigma(X_{\text{max}})$ beyond 50 EeV. A comprehensive discussion of the analysis, including the technical details of the novel X_{max} reconstruction, is given in an accompanying publication [26].

METHODOLOGY

In the past two decades, our understanding of UHECRs has grown enormously due to the construction of the Pierre Auger Observatory [27] and the Telescope Array Project [28]. The Pierre Auger Observatory is the world's largest UHECR instrument and features a hybrid detection technique, being a combination of surface detectors and fluorescence telescopes, to measure cosmic-ray-induced air showers. In total, 1660 water-Cherenkov detectors, spanning 3000 km², are arranged in a triangular 1500-meter-grid and form the Surface Detector (SD) — the centerpiece of the Observatory with a duty cycle close to 100%. The SD is overlooked by 24 telescopes located at four sites that form the Fluorescence Detector (FD) of the Observatory. Additionally, three high-elevation telescopes overlook an infilled array of 61 stations with 750 m spacing that enable measurements below 3 EeV. The FD detects the emitted fluorescence light from air molecules excited by the particle shower penetrating the atmosphere. The requirement for dark and moonless nights limits the duty cycle of the FD to about 15%.

The typical size of an air-shower footprint with E > 10 EeV amounts to tens of km², and it usually triggers more than ten stations of the SD. Each of these stations is equipped with three photomultiplier tubes (PMTs) that record the timedependent responses to shower particles digitized and sampled in steps of 25 ns. The resulting three traces are then calibrated in units of VEM (vertical equivalent muons), i.e., the average signal produced by muons traversing the detector vertically, provided by an in-situ calibration on a minute timescale using atmospheric muons. Several station-level measurements characterize each event in our analysis: the arrival time of the first particles at the respective station and, for each PMT, a trace of 3 µs time length (120 time steps) containing the signal.

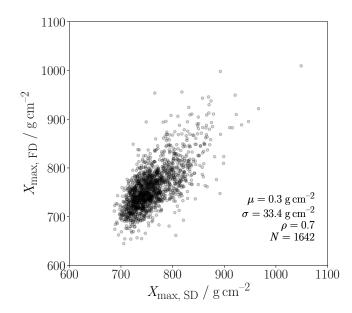


FIG. 1: Application of the DNN to hybrid data. Correlation between fluorescence observations of the FD and DNN predictions using SD data after calibration. The 1642 events show a clear correlation of $\rho = 0.7$ and a bias $< 1 \text{ g cm}^{-2}$.

In this work, we use two different data sets: a hybrid data set, featuring both an FD and SD reconstruction used to calibrate the reconstruction algorithm to the X_{max} scale of the FD, and the full SD data set for performing the high-statistics measurement of $\langle X_{\text{max}} \rangle$ and $\sigma(X_{\text{max}})$. To guarantee high data quality, we apply a dedicated event selection. We only select events with an energy $E_{SD} > 3$ EeV to ensure full trigger efficiency [29]. Additionally, we require a zenith angle $\theta < 60^{\circ}$, a hexagon of working stations around the station with the largest signal [30]. Furthermore, a fiducial SD cut is applied [26] to ensure an unbiased X_{max} measurement, accepting only events inside a zenith-angle range where the absolute X_{max} reconstruction bias of the deep-learning-based algorithm is smaller than 10 gcm^{-2} . This fiducial zenith-angle range depends on energy and was derived from simulations [26]. The SD data set comprises events collected between 1 January 2004 and 31 August 2018. After selection, the data set consists of 48,824 events. For the calibration of the novel algorithm, hybrid events featuring both an FD and SD reconstruction are used. Hence, in addition to the previous cuts, a strict FD selection is applied. We accept only events with good atmospheric conditions and small uncertainties on the observed shower profile. In particular, we reject events with X_{max} reconstructed outside the telescope field of view. Since $X_{\rm max}$ depends on the primary mass, such a cut can cause a selection bias. To avoid such a bias, we apply a fiducial cut that ensures uniform acceptance for most of the X_{max} distribution [10]. 1642 hybrid events remain after selection.

Information on the shower development and the primary particle mass is encoded in the temporal structure of the recorded SD signals, i.e., the signal traces and arrival times [12, 31]. The novel X_{max} reconstruction applied in this work is based on a deep neural network (DNN) tailored to the particular situation of the SD. Two connected sub-networks form the DNN architecture to exploit the patterns of different shower components in the time-resolved particle density. For example, muons usually produce signal spikes, while the signals from each electron, positron, and photon are individually smaller and are spread out in time because of multiple scattering [32]. In the first part, the signal traces are analyzed using long-short term memory (LSTM) layers [33] on a station-bystation level. The next part of the DNN is used to exploit the spatial distribution of the signal footprint induced on the SD grid by combining the outputs of the first part. Therefore, we utilize hexagonal convolutions [34] to account for the structure of the detector grid. We additionally use residual connections that improve the training behavior and performance of deep networks [35, 36]. Finally, the DNN predicts a single event-wise value for X_{max} [26].

The DNN was trained using the simulated detector responses [38] of 400,000 showers induced by proton, helium, oxygen, and iron with energies from 1 to 160 EeV following an E^{-1} spectrum. The showers were simulated with COR-SIKA [39] using the EPOS-LHC interaction model. Realistic operation conditions of the detector are imitated [21] during training and act as a form of data augmentation [14]. Whilst training, the DNN learns to infer an event-by-event X_{max} value following the simulated X_{max} distribution using the characteristic signal patterns of representative cosmic ray masses. For more details on the algorithm, its training, and its design, we refer to Ref. [21].

During the Observatory lifetime, the shapes of the recorded SD signals have changed due to the aging of the PMTs, the electronics, the reflective Tyvek liner, and potential changes in the water transparency. The degrading of the signal shapes is monitored using the A/P (area-over-peak) observable, which is the ratio of pulse integral and pulse amplitude for a signal produced by a single muon. It is modeled as constant in the utilized simulations but decreases with time in the array [40, 41], causing a decrease in the X_{max} predictions when applied to measured data. The predictions and their aging dependence on A/P can be described and corrected via a parameterization [21, 26]. In addition, we calibrate the DNN to remove a dependence of X_{max} on the distance of the station with the largest signal to the reconstructed shower core. Furthermore, we examine seasonal and diurnal variations of the X_{max} reconstruction and correct it by fitting an oscillating function. The amplitudes found are below 2 g cm^{-2} .

Finally, we use hybrid data to calibrate the SD-based DNN algorithm to the FD X_{max} scale. Since UHECRs feature energies several orders of magnitude above what can be reached with human-built accelerators, air-shower simulations need to make use of extrapolations of accelerator data. The phenomenological modeling and the extrapolations used differ for each hadronic interaction model. As fluorescence telescopes directly observe X_{max} , they offer the possibility of removing the dependence of the SD-based algorithm on the

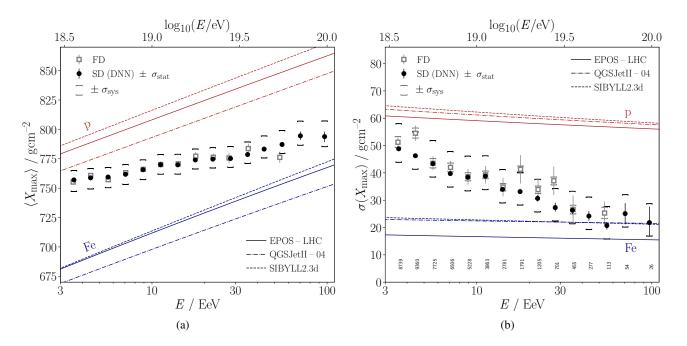


FIG. 2: Energy evolution of (a) the mean depth of shower maximum $\langle X_{max} \rangle$ and (b) the fluctuations of shower maximum $\sigma(X_{max})$ as determined using the FD reconstruction (grey open squares) [37] and the SD-based DNN predictions (black circles). Red (blue) lines indicate expectations for a pure proton (iron) composition for various hadronic models. The number of events in each bin is indicated in panel (b).

particular interaction model and significantly reduce the systematic uncertainties of the $\langle X_{\text{max}} \rangle$ measurement. By studying the difference between the DNN predictions and the FD observations, we observe an offset of (-31.7 ± 0.7) g cm⁻² compatible within uncertainties to be independent of energy $(\Delta X_{\text{max}} < 6 \text{ g cm}^{-2} \text{ decade}^{-1})$, as determined by a fit. The observed offset is larger than the expected differences by up to -15 g cm^{-2} from studies using various hadronic interaction models [21, 26]. This indicates that the current generation of hadronic interaction models may not describe the measured data entirely, which is consistent with previous analyses that suggest inadequacies in the description of muon profiles [12, 31, 42, 43], as well as the longitudinal profiles in general [44].

In Figure 1, the correlation between the X_{max} reconstruction of the DNN and the FD is shown after calibration. The observed Pearson correlation of $\rho = 0.7 \pm 0.03$ is a significant improvement compared to previous analyses that utilize the SD data to extract mass-dependent information [12]. The correlation and resolution of the DNN are in excellent agreement with simulation studies verifying the reconstruction of the DNN and indicating that these fluctuations are well modeled. This can be expected, as the shower fluctuations are significantly driven by the fluctuations of the first interaction [45], which is relatively similar across hadronic interaction models, and additionally, the relative fluctuations of the number of muons seem to be properly modeled [43].

RESULTS AND DISCUSSION

To investigate the evolution of the UHECR mass composition, we study the first and second moments of the X_{max} distributions measured by the SD as a function of energy E. We use bins of $\Delta \log_{10}(E/\text{EeV}) = 0.1$ and an integral bin beyond $10^{19.9}$ eV. The energy evolution of the measurement of $\langle X_{\text{max}} \rangle$ and $\sigma(X_{\text{max}})$ is shown in Figure 2. The grey open squares denote FD measurements [37] of the same data-taking period, and black circles the SD-based DNN measurement of this work, extending the X_{max} measurements to the highest energies. Whereas vertical bars indicate statistical uncertainties obtained via bootstrapping, brackets denote systematic uncertainties. The red (blue) lines mark predictions [46] from three hadronic interaction models [47-49] for a pure proton (iron) composition. The systematic uncertainties of $\langle X_{
m max}
angle$ range from 9 $g cm^{-2}$ to 13 $g cm^{-2}$ and are dominated by the hybrid calibration and the uncertainty of the FD X_{max} scale. The systematic uncertainties of the $\sigma(X_{\text{max}})$ measurement are dominated by the composition bias of the energy measurement and the interaction-model bias of the DNN, and decrease with increasing energy. This corresponding bias was conservatively estimated using a simulation study with various realistic composition scenarios, the measured UHECR energy spectrum [4], and by considering systematic uncertainties on the reconstruction [26].

The $\langle X_{\text{max}} \rangle$ measured with the SD shows excellent agreement with FD observations as shown in Figure 2a. The measurement shows a transition from a relatively light to a heavier composition, confirming the observation of previous

analyses [12, 37, 50–52] and extending our measurements to 100 EeV. As shown in Figure 2b, with rising energy, the fluctuations diminish and agree well with previous FD measurements. The observation of decreasing $\sigma(X_{\text{max}})$ implies that besides becoming heavier, the mass composition also has to be rather pure. This yields a consistent interpretation of the primary UHECR composition when combined with measurements of $\langle X_{\text{max}} \rangle$. The small fluctuations disfavor a substantial fraction of light particles at the highest energies and, at the same time, indicate that the observed suppression in the energy spectrum cannot be entirely ascribed to effects of extragalactic propagation [8, 9].

A change in the composition of the primary mass can be studied by investigating the *elongation rate*:

$$D_{10} \stackrel{\circ}{=} \frac{\mathrm{d}\langle X_{\mathrm{max}}\rangle}{\mathrm{d}\log_{10}E} = \hat{D}_{10} \left(1 - \frac{\mathrm{d}\langle \mathrm{ln}A\rangle}{\mathrm{d}\,\mathrm{ln}\,E}\right),\,$$

defined by the change of $\langle X_{max} \rangle$ in one decade of energy and comparing it to an expected elongation rate \hat{D}_{10} obtained using simulations, which is to a good approximation universal across all primary masses A and hadronic interaction models and ranges from 55 to 60 $gcm^{-2}decade^{-1}$. A linear fit with a constant elongation rate yields $D_{10} = 24.1 \pm$ $1.2 \text{ g cm}^{-2} \text{decade}^{-1}$, in good agreement with the FD measurements in this energy range $((26 \pm 2) \text{ g cm}^{-2} \text{ decade}^{-1})$, but does not describe well our data with $\chi^2/dof = 46.7/13$. Due to the significant increase in statistics, we find evidence for a distinctive structure in the transition towards a heavier composition. We study the energy dependence of $\langle X_{\text{max}} \rangle$ using a function piece-wise linear in log(E/eV) with three breaks. The observed elongation rate model, shown as a red line in the top panel of Figure 3, features three breaks $(\chi^2/dof = 10.4/7)$ at which the elongation rate changes. Using Wilks' theorem, we compared this model with the null hypothesis of a constant elongation rate and found that we can reject the constant elongation rate model at a statistical significance of 4.6 σ . Considering energy-dependent systematic uncertainties, the significance level for rejecting a constant elongation rate reduces to 4.4 σ . We furthermore studied the compatibility of the FD data with our new elongation rate model and observed a good agreement ($\chi^2/dof = 12.8/12$). The null hypothesis of a model describing only two breaks at lower energies (E_1, E_2) , positioned close to the ankle and instep, can be rejected at a statistical significance level of 3.3σ using the found elongation rate model. The rejection of the two-break model hypothesis shows a stronger dependence on systematic uncertainties due to the low statistics in the hybrid data set at high energies (E > 30 EeV) used for investigating the energy dependence of the DNN calibration. A single-break model can be rejected with a significance of 4.4σ and consistently remains above the 3σ level when including systematics.

The fitted parameters of the model with three breaks are summarized in Table I together with the positions of the energy of spectrum features measured using the SD and the infill array with 750 m spacing. As shown as a continuous red line in the top panel of Figure 3, the found breaks in the evolution

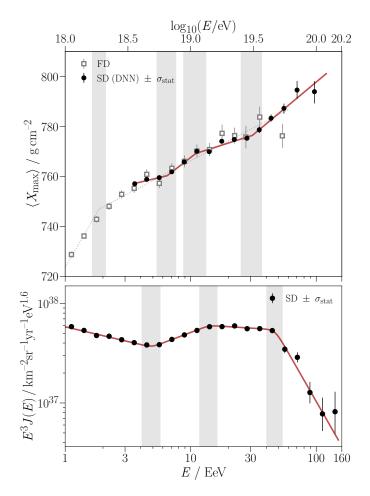


FIG. 3: Positions of breaks in the elongation rate compared to the features identified in the energy spectrum. Top: Evolution of $\langle X_{max} \rangle$ as a function of energy for the SD (black) and the FD (grey) [37]. The red line indicates the elongation model found using the SD, and the dotted grey line using the FD. Bottom: Combined energy spectrum [53] as measured using the SD 1500 m array and the low energy 750 m infill array of the Observatory. Grey regions indicate the uncertainties in the energy of the found breaks in the $\langle X_{max} \rangle$ evolution and features in the energy spectrum.

of $\langle X_{\text{max}} \rangle$ are observed close to the ankle, instep, and suppression features of the energy spectrum [53], shown in the bottom panel of Figure 3. The hatched grey regions denote statistical and systematic uncertainties of the position of the features. Note that distinct features do not have to emerge at similar energies for an astrophysical interpretation of the energy spectrum and its composition. For example, the break in the elongation rate observed using the FD of the Observatory around 2 EeV [51], shown as a dotted grey line in the top panel of Figure 3, is physically interpreted [8, 9, 54] in association with the ankle, which has been discovered at 5 EeV.

Interestingly, the composition model discussed in Ref. [9] (Fig. 3 and Fig. 6), derived by taking into account astrophysical scenarios, including extragalactic propagation and fitting the energy spectrum measured by the SD and the X_{max} distribution observed by the FD, predicts three breaks at positions

TABLE I: Best-fit parameters with statistical and systematic uncertainties for the identified elongation model that features three changes at energies (E_1, E_2, E_3) in the elongation rate (D_0, D_1, D_2, D_3) and an offset *b* of $\langle X_{\text{max}} \rangle$ at 1 EeV. The positions of the features of the energy spectrum [53] are also given.

parameter	3-break model	energy spectrum
val $\pm \sigma_{\rm stat} \pm \sigma_{\rm sys}$	$val\pm\sigma_{stat}\pm\sigma_{sys}$	$val\pm\sigma_{stat}\pm\sigma_{sys}$
$b/g \text{ cm}^{-2}$	$750.5 \pm 3 \pm 13$	
D_0 / g cm ⁻² decade ⁻¹	$12\pm5\pm6$	
E_1 / EeV	$6.5 \pm 0.6 \pm 1$	$4.9 \pm 0.1 \pm 0.8$
D_1 / g cm ⁻² decade ⁻¹	$39\pm5\pm14$	
E_2 / EeV	$11\pm2\pm1$	$14\pm1\pm2$
D_2 / g cm ⁻² decade ⁻¹	$16\pm3\pm6$	
E_3 / EeV	$31\pm5\pm3$	$47\pm3\pm6$
D_3 / g cm ⁻² decade ⁻¹	$42\pm9\pm12$	

matching our findings. These indications reinforce our results and will be further investigated to extract information on their astrophysical origin.

We also studied the evolution of $\sigma(X_{\text{max}})$ (see Figure 2b) to identify a potentially similar underlying structure. We observe a decrease in fluctuations, while the elongation rate implies a change towards a heavier composition. Consistently, we find no substantial change in the fluctuations $\sigma(X_{\text{max}})$ at the regions - between the ankle and the instep, and above the suppression — where the elongation rate of $\langle X_{\text{max}} \rangle$ is closer to that of a constant composition. While being compatible with the data ($\chi/ndf = 10.3/10$), a model featuring three breaks at positions fixed to those found in the elongation rate is statistically not significant. Using such a model, a linear decrease in $\sigma(X_{\text{max}})$ can be rejected at a 2.2 σ significance level only. Note that changes in the primary mass composition are not reflected in the same way in the energy evolution of $\langle X_{\text{max}} \rangle$ and $\sigma(X_{\text{max}})$ [11]. A simple transition between two primary species at a constant rate corresponds to a linear dependence of $\langle X_{\max} \rangle$ on log(*E*) but to a non-linear behavior of $\sigma(X_{\max})$, for which, thus, the application of a broken-line model is inappropriate. For the evolution involving a larger number of primary species with unknown proportions, a specific model for the interpretation of $\sigma(X_{\text{max}})$ cannot be defined. Hence, more sophisticated investigations are needed. Detailed studies on the astrophysical origin of the features in the mass composition and the energy spectrum are ongoing and will, jointly with the AugerPrime upgrade, offer new insights into the nature of UHECRs.

SUMMARY

We have performed a measurement of $\langle X_{max} \rangle$ and $\sigma(X_{max})$ for cosmic rays with energies between 3 and 100 EeV to investigate their mass composition. The method relies on the timedependent signals recorded by the SD of the Pierre Auger Observatory. After training our deep learning model on simulated SD data, we used measured hybrid data to crosscheck and cross-calibrate our algorithm using the FD of the Observatory to remove mismatches between simulations and measured data. With the calibrated method, we obtained a 10fold increase in the size of the X_{max} data set for E > 5 EeV compared to the FD measurements and found a consistent picture of the $\langle X_{\text{max}} \rangle$ and $\sigma(X_{\text{max}})$ measurements. At lower energies, our measurements are in excellent agreement with fluorescence observations, indicating a light and mixed mass composition. At the highest, so far inaccessible, energies, we report a purer and heavier composition, confirming the trend indicated by the FD data. The observation of small fluctuations in X_{max} beyond 50 EeV further excludes a significant fraction of light nuclei at the highest energies. Due to the substantial rise in statistics, we have found evidence at a level of 4.4 σ for a characteristic structure in the evolution of the mass composition beyond a constant elongation rate. The model describing our data best features three breaks. Interestingly, the identified breaks in the elongation rate model are observed to be in proximity of the ankle, instep, and suppression features in the energy spectrum, where changes in the spectral index have been reported [4]. A structure, while not statistically significant, is visible in $\sigma(X_{\text{max}})$, which could suggest breaks at similar energies. More statistics are needed to study the nature of the identified breaks and, particularly, investigate the existence of the third break at the highest energies, where the statistics are decreasing and systematic uncertainties are increasing. We have demonstrated the significant potential of applying deep neural networks to astroparticle physics, particularly in the analysis of low-level data. Our approach comprises a detailed study of systematic uncertainties, including the cross-calibration with a complementary detector, highlighting the importance of an independent data set for calibration and validation of these powerful algorithms.

The Pierre Auger Observatory is now being upgraded, which includes the deployment of scintillators and radio antennas on top of each SD station. The new detectors, combined with the emerging capabilities of machine-learningbased algorithms, offer unique prospects for accurate composition studies [55, 56] and increase our understanding of cosmic rays at ultra-high energies.

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