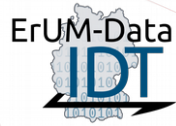


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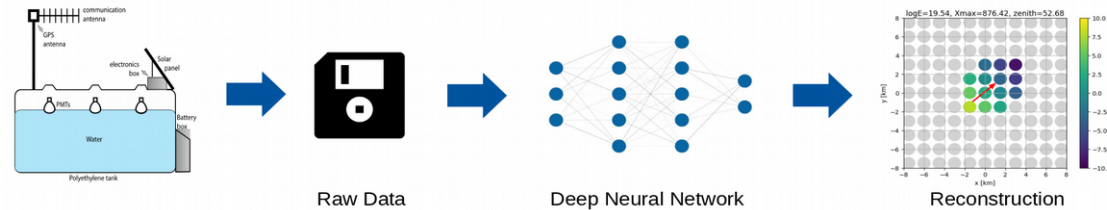


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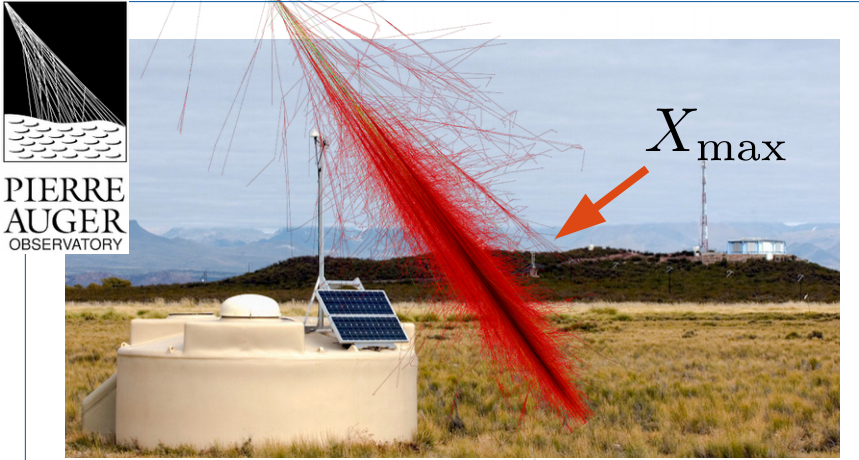
Cosmic Ray Indirect – Contribution 915

Event-by-event reconstruction of X_{\max} with the Surface Detector of the Pierre Auger Observatory using deep learning



Jonas Glombitza on behalf of the Pierre Auger Collaboration

Ultra-high-energy cosmic rays (UHECRs)



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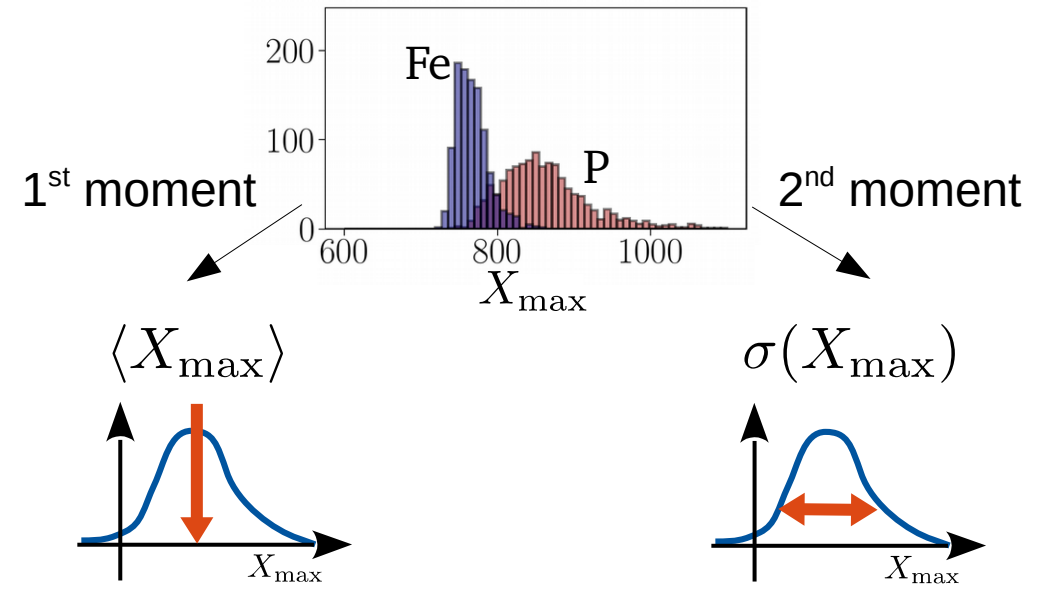
X_{\max}

The Pierre Auger Observatory

- world's largest observatory to study ultra-high-energy cosmic rays
- hybrid detection of air showers
 - 1,660 water-Cherenkov detectors
 - 27 fluorescence telescopes

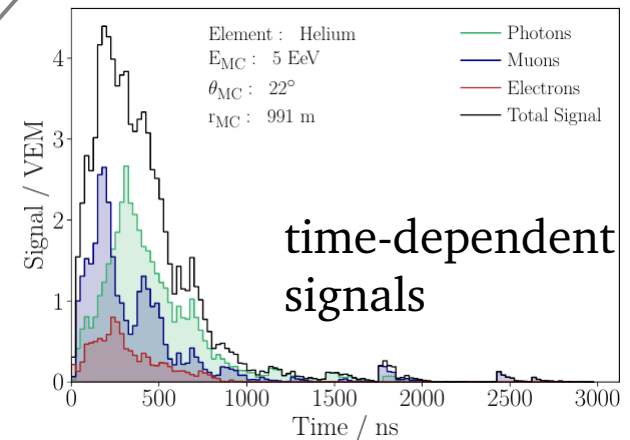
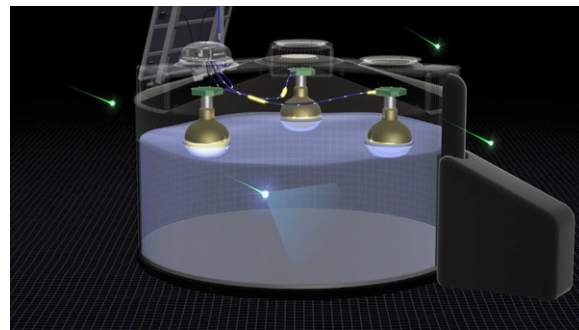
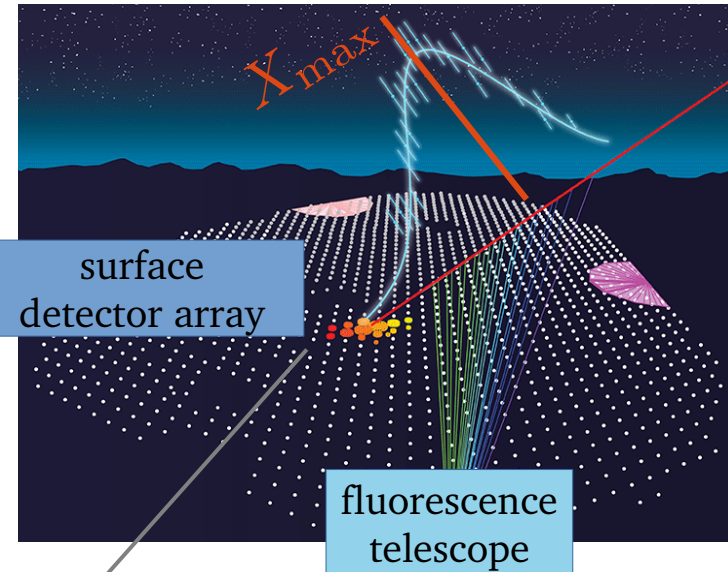
Mass composition of UHECRs

- currently: most precise mass estimate by reconstructing shower maximum X_{\max}
- determine composition by studying the measured X_{\max} distributions



Reconstructing the shower maximum X_{\max}

- precise observation of X_{\max} using fluorescence telescopes
 - observations confined to dark nights (15% duty cycle)
- challenging to measure with surface detector (SD)
 - no direct observation of X_{\max} possible
 - measured time-dependent signals encode information about air-shower development
- use deep learning to estimate X_{\max}



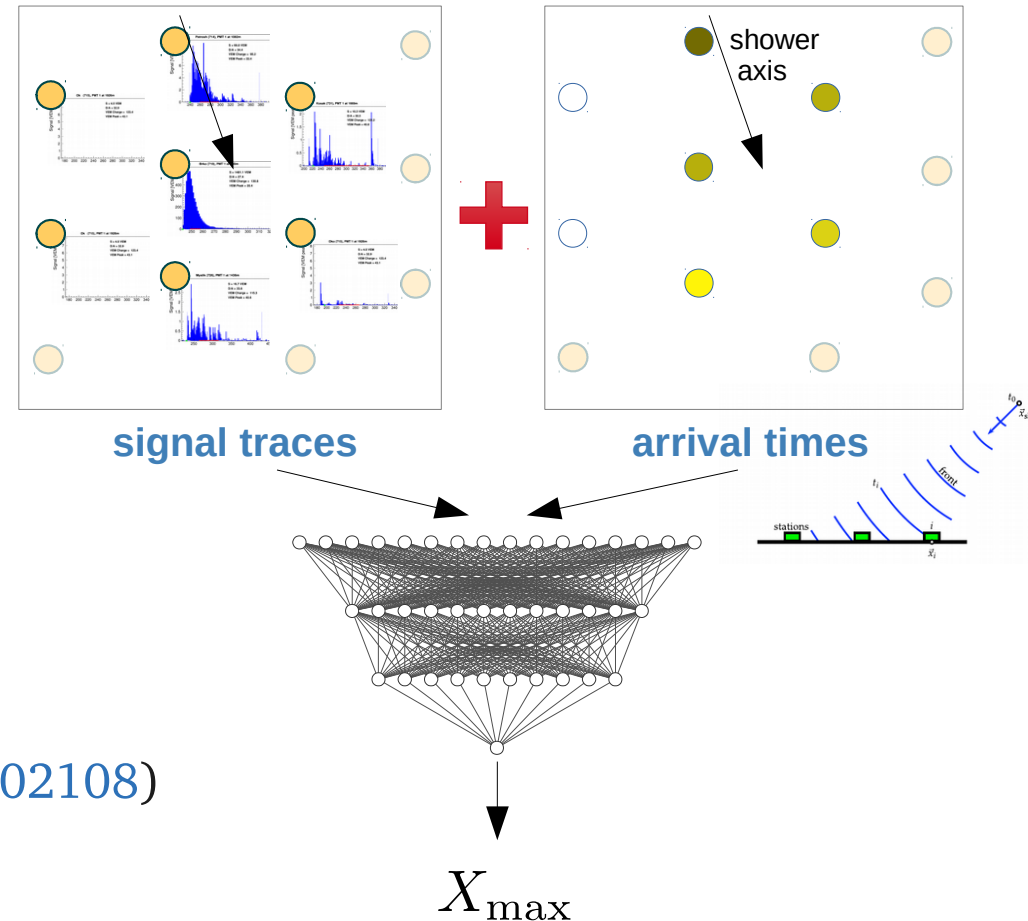
Deep-learning based reconstruction of X_{\max} using the Surface Detector

SD provides measurement of:

- time-dependent signals
- arrival time of first shower particles
- use deep neural network to reconstruct shower maximum

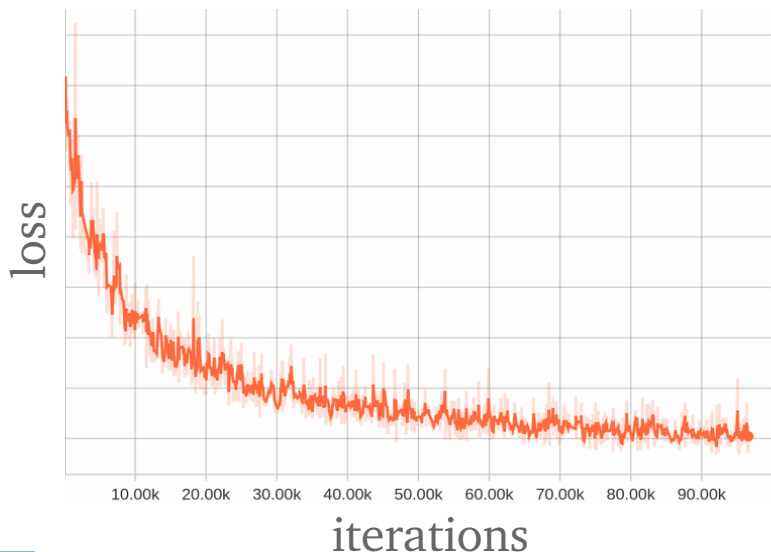
Reconstruction strategy:

- (1) process time-dependent signal traces using recurrent networks (LSTMs)
- (2) process shower footprint
 - exploit symmetry of the SD using hexagonal convolutions ([ArXiv/1803.02108](https://arxiv.org/abs/1803.02108))



Network training

- ~ 1.5 million parameters
- training on GPU ~ 1-2 days
- *mimic different detector states*
 - ♦ *broken stations, broken PMTs*
 - ♦ *various electronic saturations*



Xmax reconstruction using deep learning
Glombitza | RWTH Aachen | ICRC 2021

	Epos LHC
# Showers	800,000
Training	700,000
Validation	10,000
Test	90,000
Energy $\log_{10}(E/\text{ev})$	18.0 – 20.2
Spectrum	E^{-1}
Composition	25% proton 25% helium 25% oxygen 25% iron
Zenith	0 – 65°

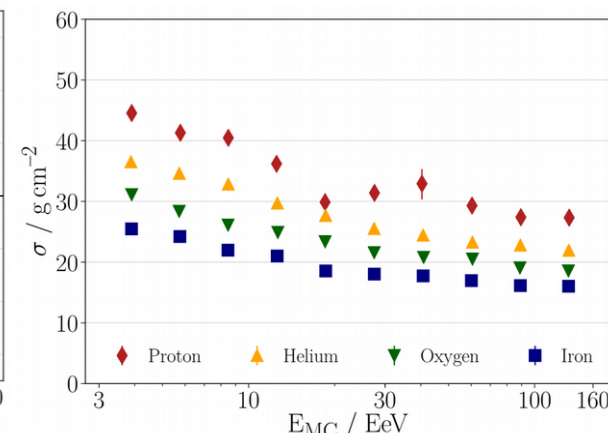
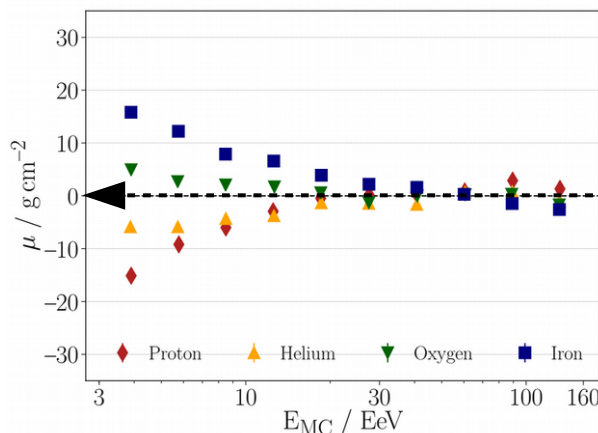
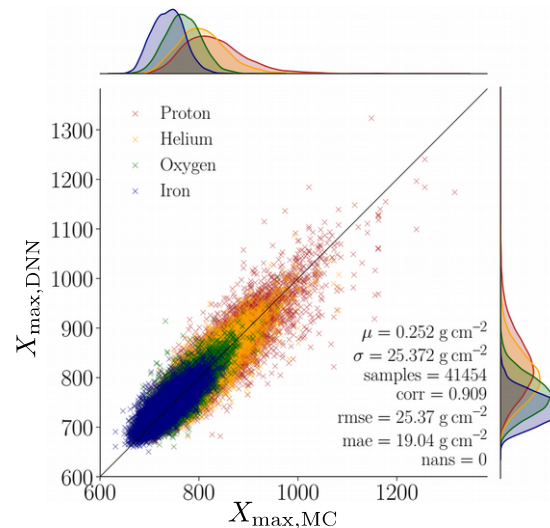
DNN trained on EPOS-LHC only!

Validate on EPOS-LHC, QGSJetII-04, Sibyll2.3c

Evaluation – EPOS-LHC

DNN trained using EPOS-LHC

- evaluation using EPOS-LHC
- performance improves with energy
- above 10 to 20 EeV
 - ♦ bias vanishes
 - ♦ proton resolution $\sim 30 \text{ g/cm}^2$
 - ♦ iron resolution $\sim 20 \text{ g/cm}^2$
- averaged among compositions
 - ♦ overall bias $\sim 0 \text{ g/cm}^2$



Evaluation - additional interaction models

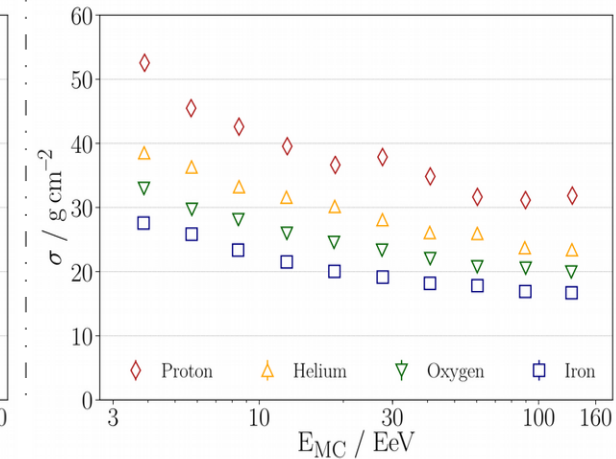
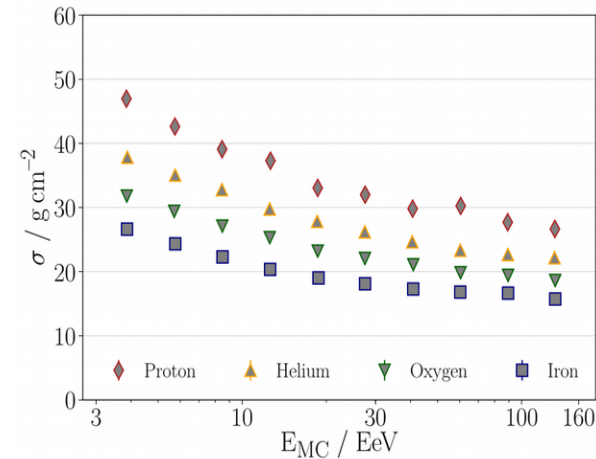
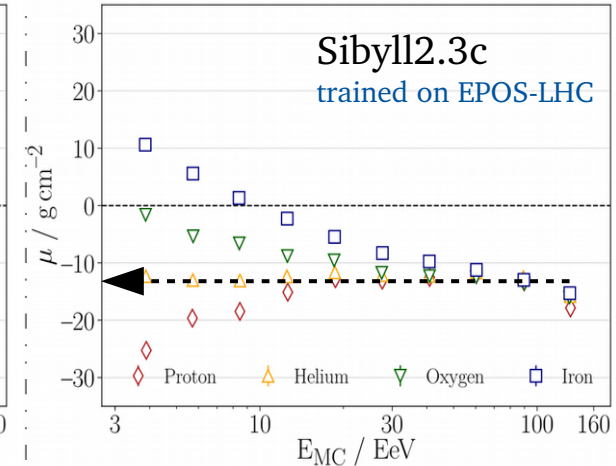
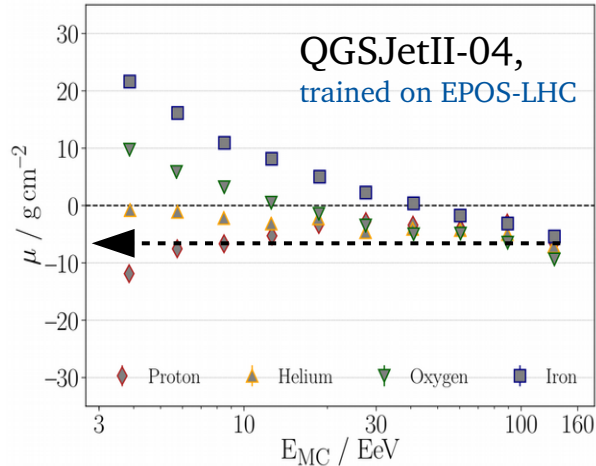
DNN trained using EPOS-LHC

Evaluation using:

- QGSJET-II.04
- SIBYLL2.3c

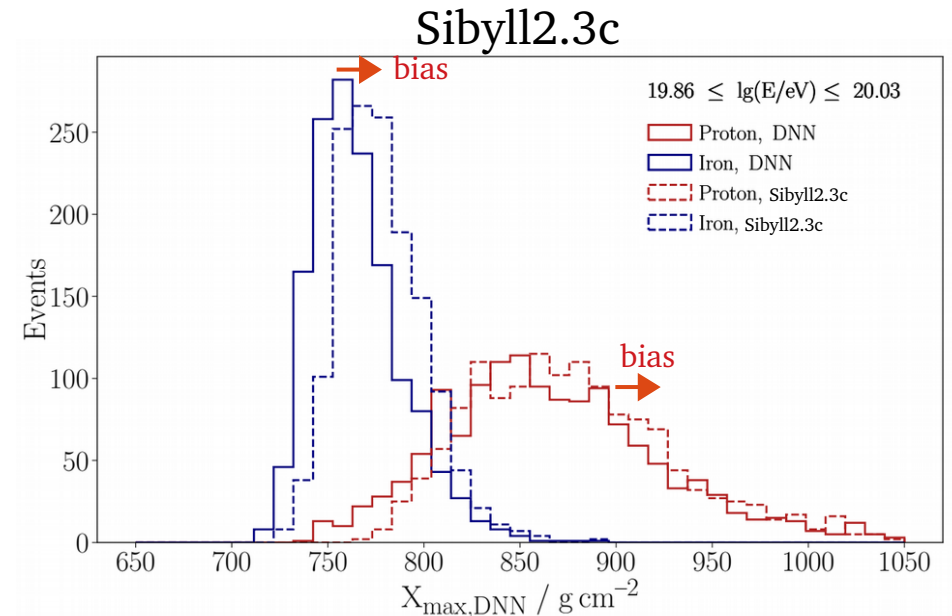
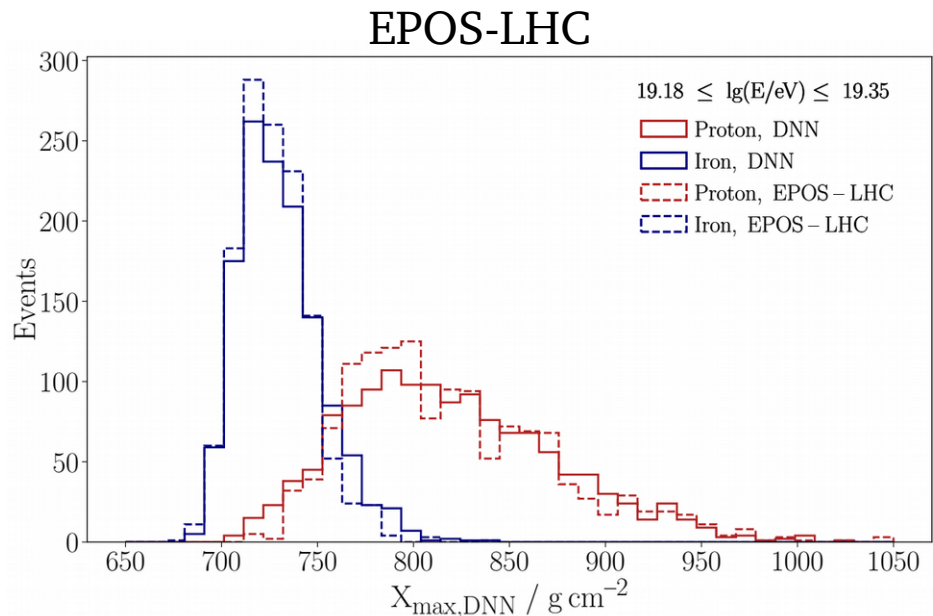
- similar resolution
 - interaction model independent

- bias different
 - ♦ mean negative for both models
 - Xmax scale of the DNN depends on interaction model



Reconstructed X_{\max} distributions

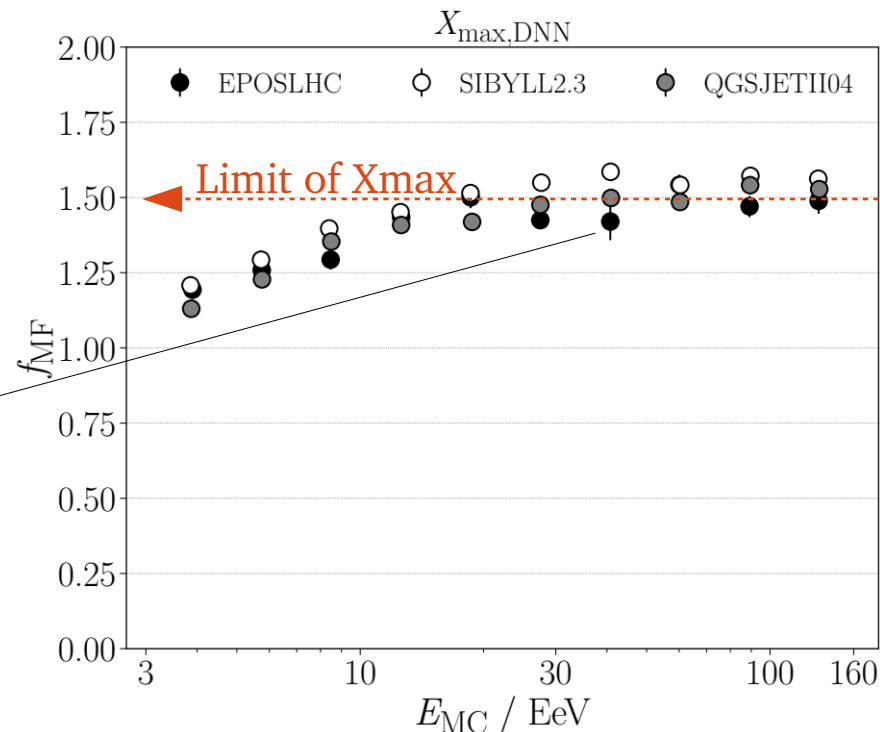
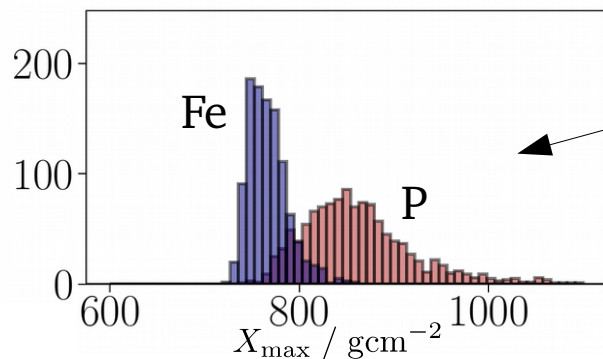
- check reconstructed X_{\max} distributions for various hadronic models
- absolute bias clearly visible, as expected (Sibyll2.3 = -15 g/cm^2)
- overall shape reconstructed correctly
- calibration to X_{\max} -scale of the FD needed for measuring X_{\max} distributions

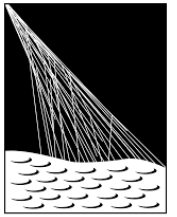


Merit factor for discriminating between proton and iron

$$f_{\text{MF}} = \frac{|\langle X_{\text{max,P}} \rangle - \langle X_{\text{max,Fe}} \rangle|}{\sqrt{\sigma^2(X_{\text{max,P}}) + \sigma^2(X_{\text{max,Fe}})}}$$

- merit factor of simulated $X_{\text{max,MC}} \sim 1.5$
- DNN merit factor increases with energy
 - ♦ above 10 EeV, merit factor = 1.5
 - ♦ good separation for all interaction models





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Application to hybrid data

Jonas Glombitza on behalf of the Pierre Auger Collaboration

Calibration to account for detector ageing

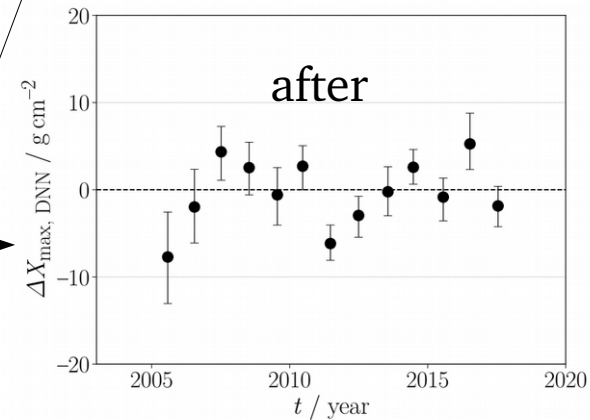
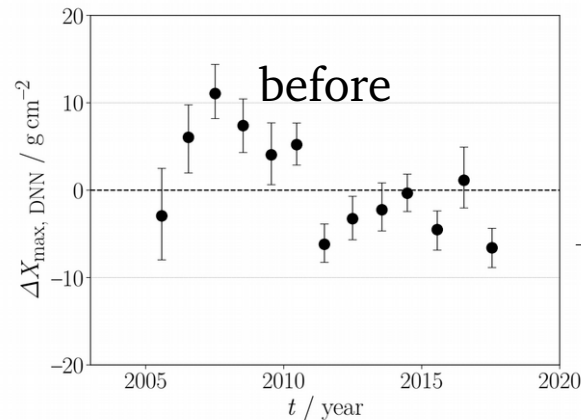
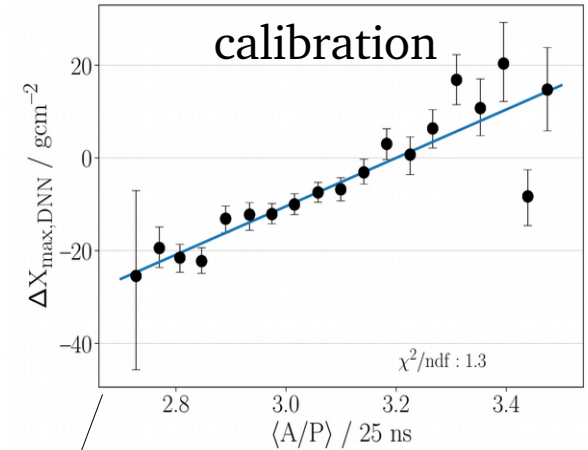


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Ageing of detector stations slightly affects signal shapes

- width of muon signals decays over time
- detector monitoring records evolution of $\langle A/P \rangle$
 - ♦ ratio between signal area and signal peak
- perform linear calibration with respect to $\langle A/P \rangle$
 - ♦ ageing effect removed

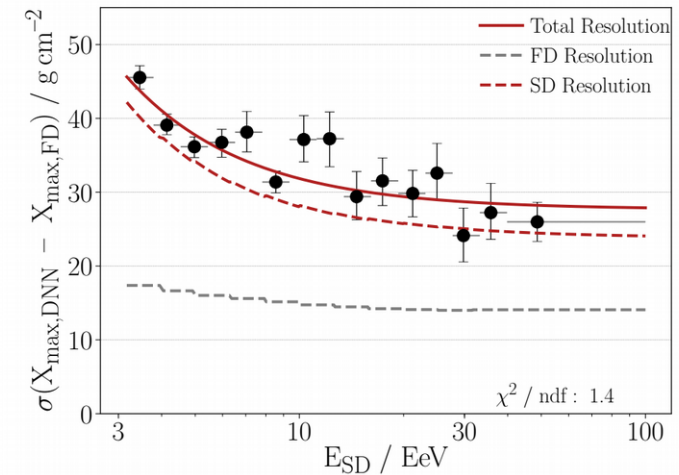
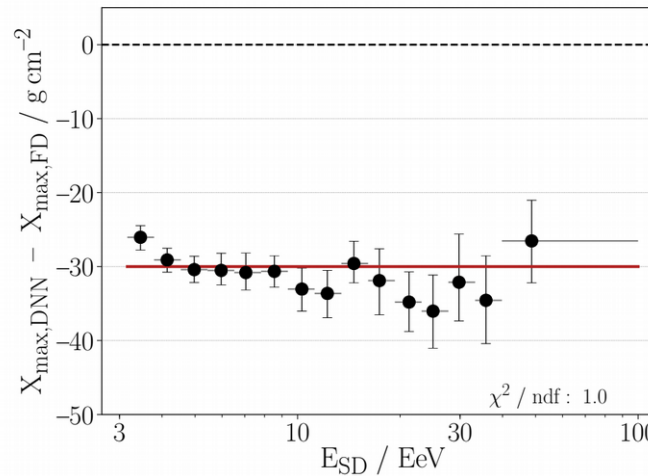
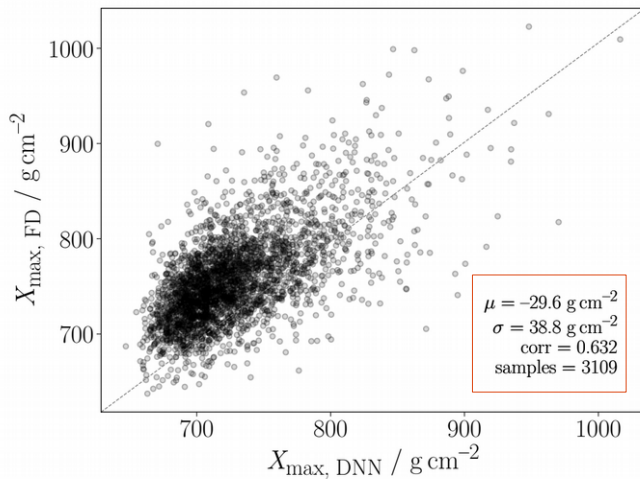


Long Term Performance of the Pierre Auger Observatory
<https://doi.org/10.22323/1.358.0222>

Application to measured hybrid data

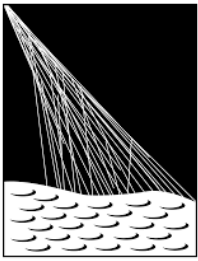
Use hybrid events to perform calibration to the FD X_{\max} scale

- **correlation 0.63** (0.61 when corrected for elong. rate)
- **resolution matches** expectations, at 10 EeV below 30 g/cm^2
- **-30 g/cm^2 bias** (hadronic models, detector simulation)
 - ♦ independent of energy \rightarrow perform calibration



Deep-learning based reconstruction of X_{\max} using the SD ([ArXiv/2101.02946](https://arxiv.org/abs/2101.02946))

- reconstruction extensively studied on various hadronic interaction models
 - ◊ X_{\max} -bias depends on interaction model used
- event-by-event resolution independent of interaction model
- performance validated on golden hybrids
 - ◊ bias of -30 g/cm^2 observed \rightarrow DNN calibrated to the FD X_{\max} scale
 - ◊ observed resolution meets expectations from simulation studies
- very promising results to measure X_{\max} -distributions to the highest energies
 - \rightarrow new insights into the UHECR composition at highest energies!
 - \rightarrow new prospects for analyses requiring event-by-event estimation of primary mass
- AugerPrime will allow for additional cross-checks and improvements



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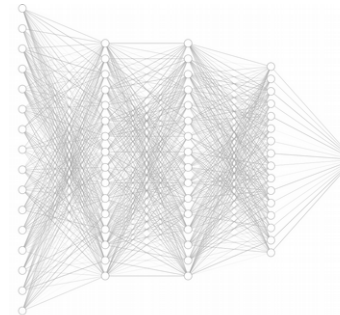
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Event-by-event reconstruction of X_{\max} with the Surface Detector of the Pierre Auger Observatory using deep learning

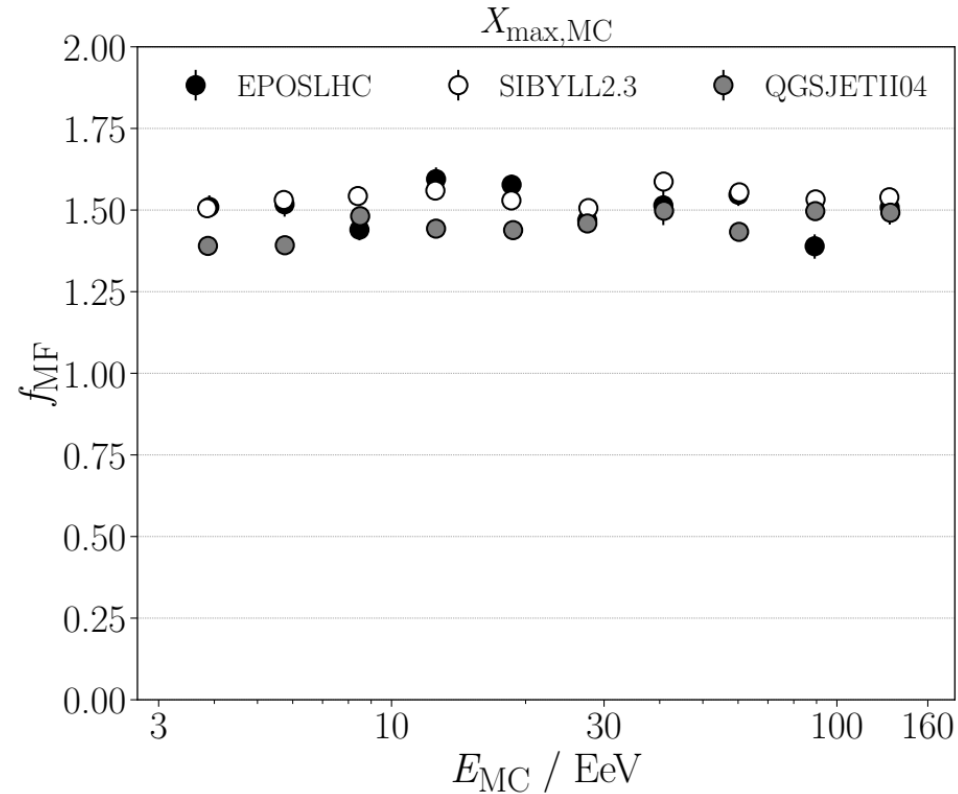
Backup

Jonas Glombitza on behalf of the Pierre Auger Collaboration

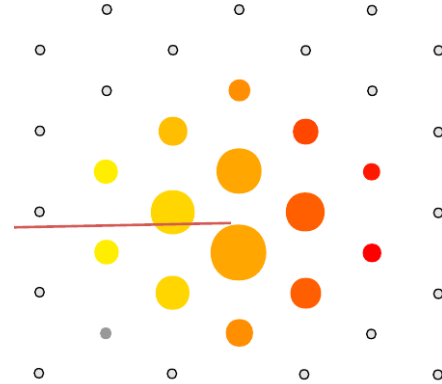
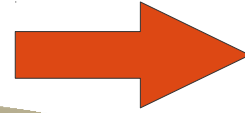
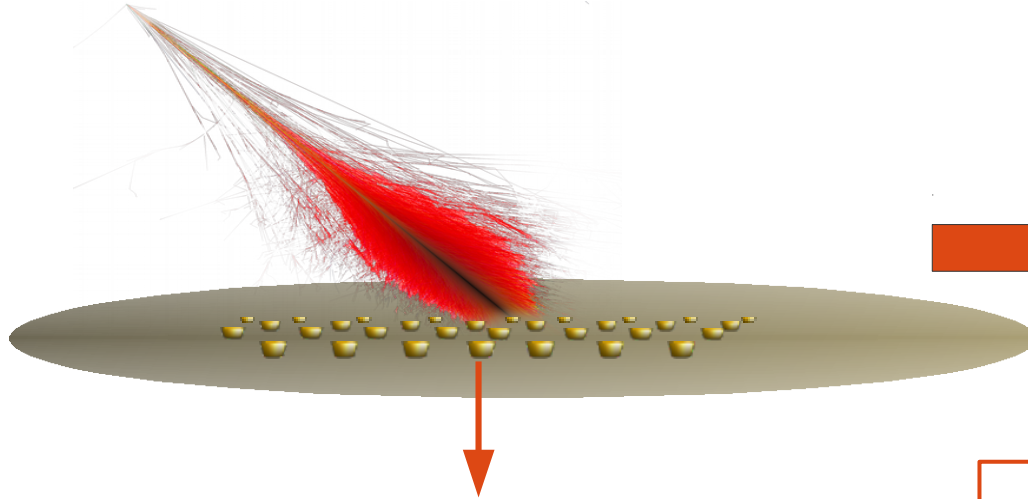


- Data from 01.01.2004 till 31.8.2018
 - **SD selection**
 - ♦ $\text{Log}_{10}(E/\text{eV}) > 18.5$
 - ♦ Zenith angles below 60 degrees
 - ♦ Only events in which the station with the highest signal is surrounded by a hexagon of operating stations
 - ♦ include events with saturated stations
 - **FD selection**
 - ♦ Same as used for the X_{max} measurements with the FD
- Full sample: 3124 events

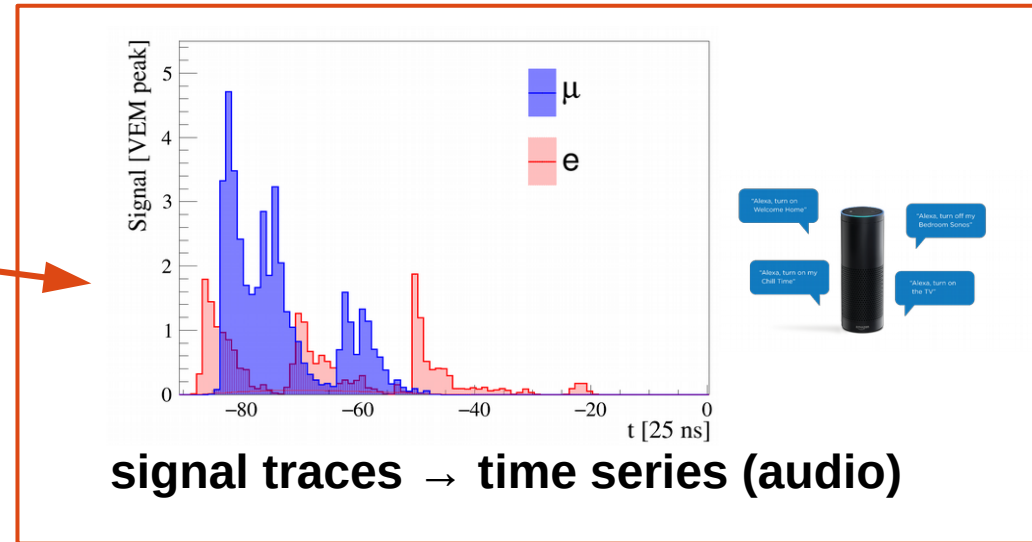
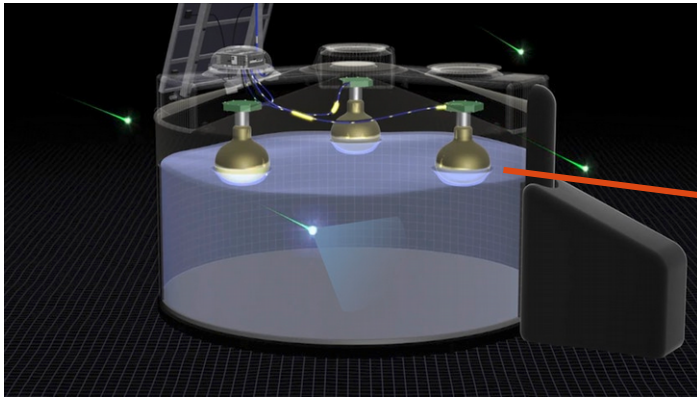
Merit factor for the simulated Xmax



Air-Shower Detection



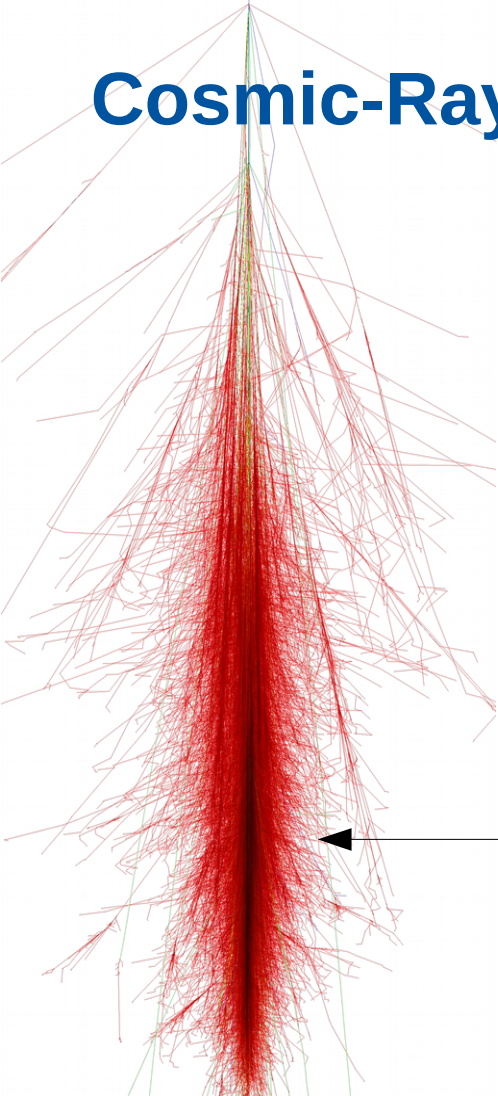
air-shower footprint



signal traces → time series (audio)

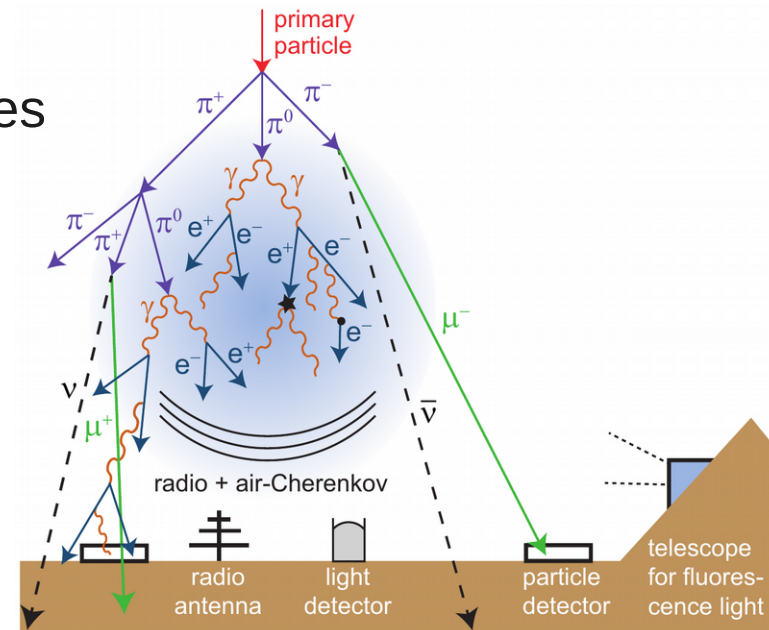
Cosmic-Ray induced Air Showers

- cosmic rays interact with Earth's atmosphere
 - induce extensive particle cascade
- particle shower reach size of several km² at Earth's surface
- particle mass determines shower structure
 - ◆ low mass, deep penetration → late maximum
 - ◆ heavy mass, early maximum
- many different detection techniques



X_{max}

shower maximum
correlates with
primary mass

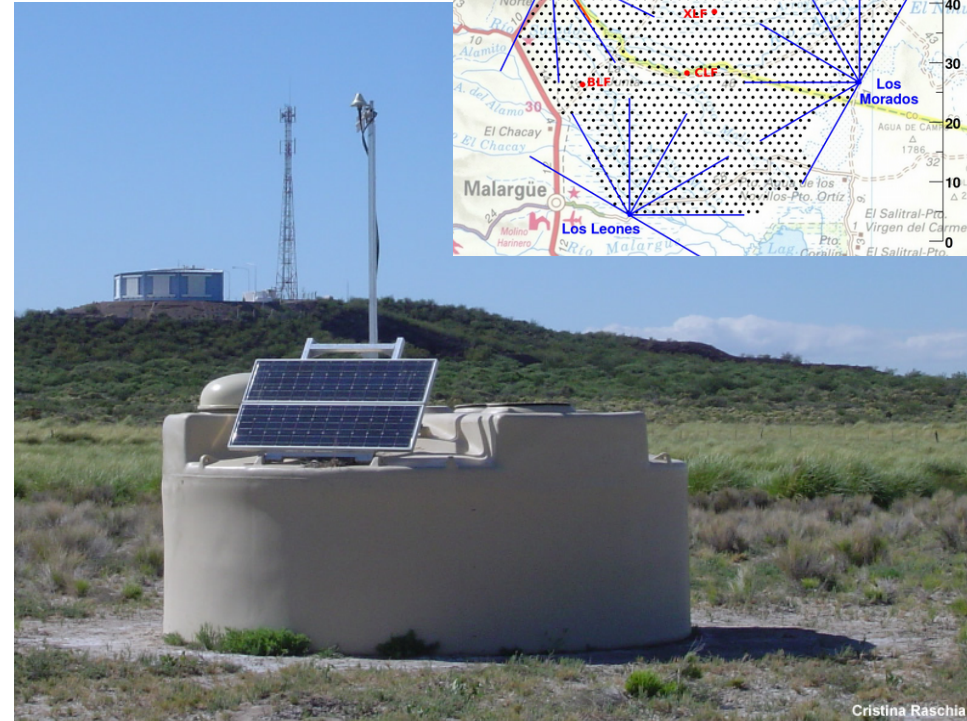


The Pierre Auger Observatory

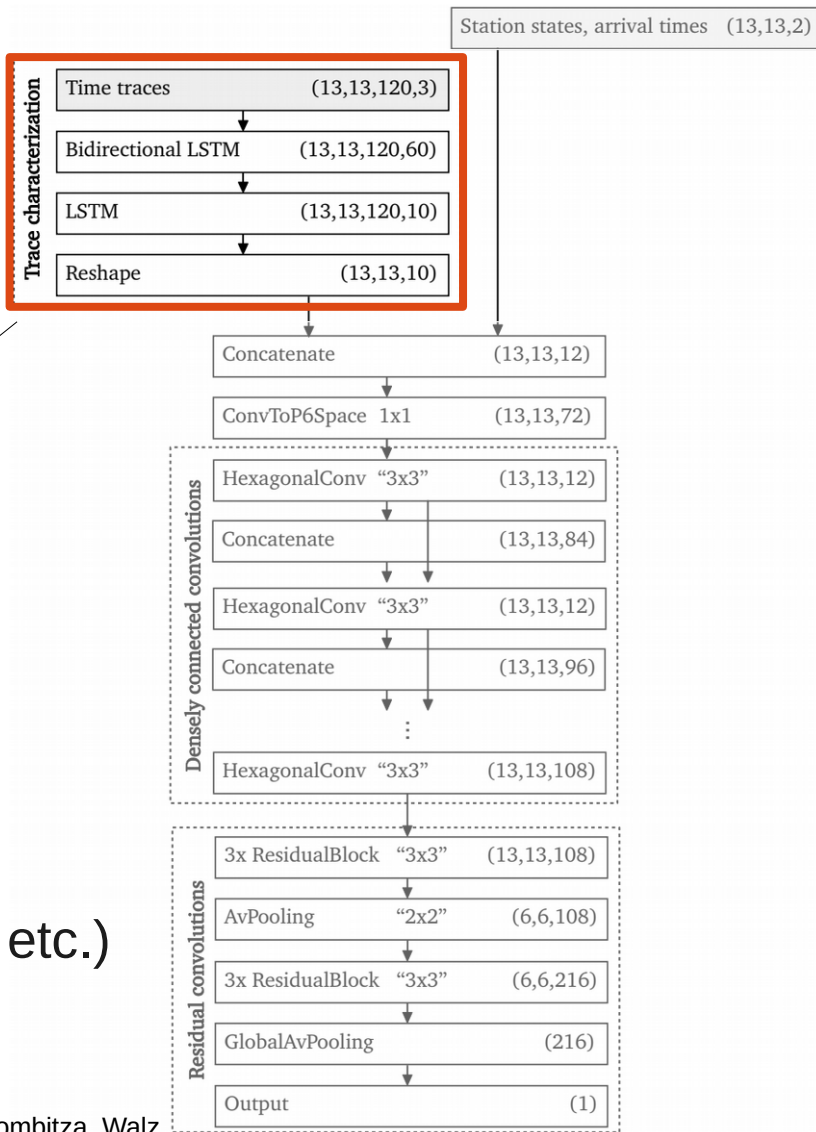
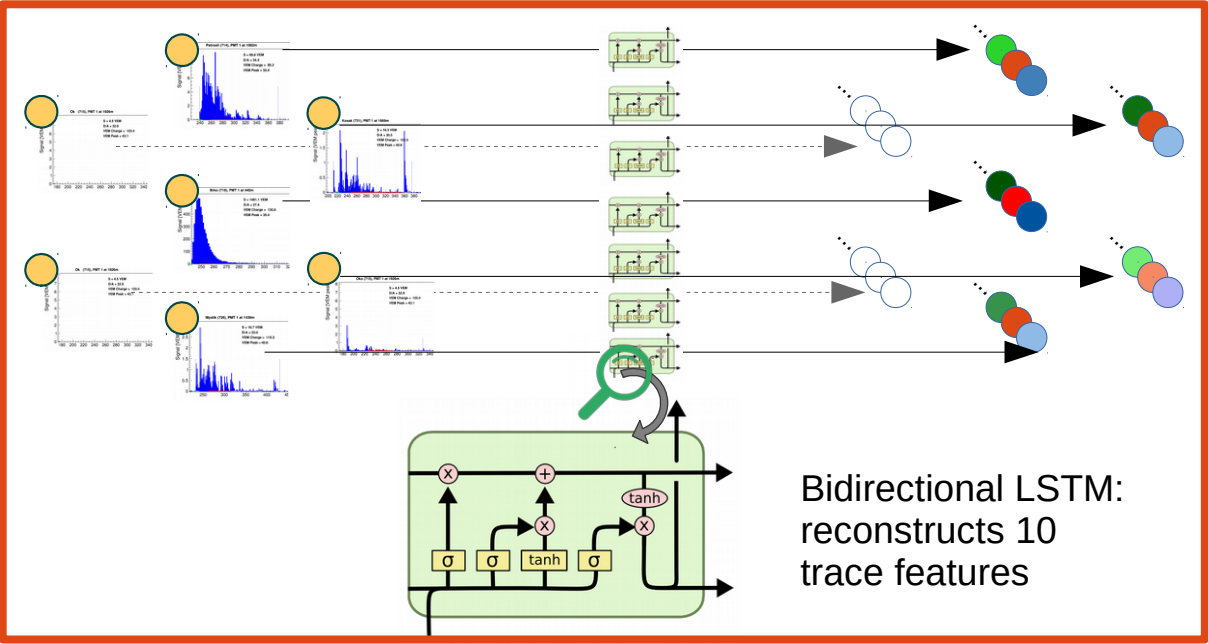
- world's largest cosmic-ray observatory
- placed in Argentina
- measure high-energetic particles
 - ♦ energy $> 10^{17}$ eV
- study composition of cosmic rays
- search for cosmic-ray origins

hybrid measurements of UHECRs

- 27 fluorescence telescopes at 4 sites
 - ♦ 15% duty cycle
- 1660 water-Cherenkov stations
 - ♦ 3000 km² array, ~100% duty cycle

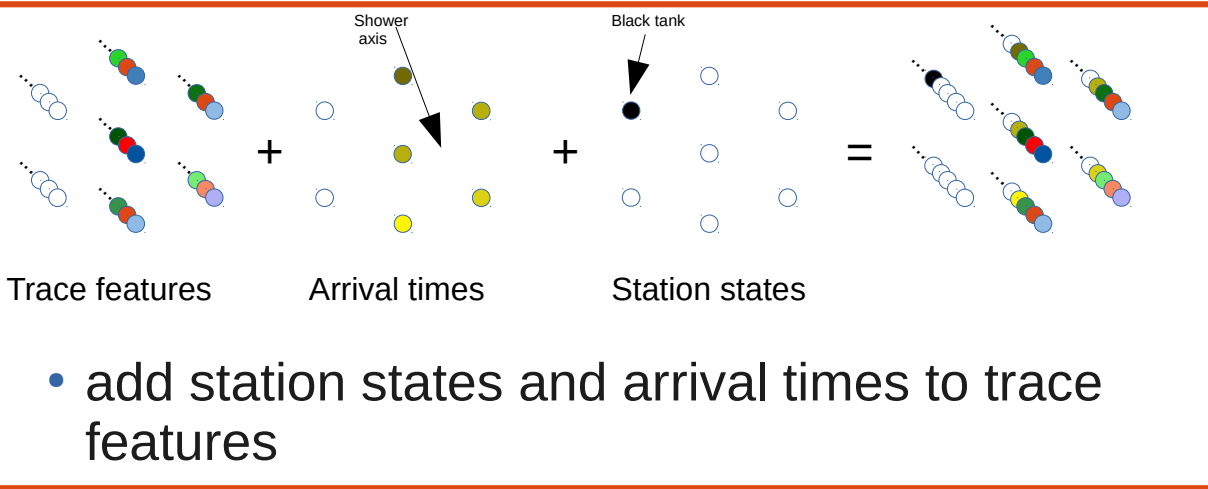


Cristina Ráschia



RNN part extracts trace features

- **same network** (same weights) for each station
 - same kind of features (same color) per station (features, e.g., rising edge, falling edge, peaks, etc.)
 - but different characteristic → different strength



- add station states and arrival times to trace features

