# Reconstruction of muon LDF using ADC and binary channels of UMD

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## UMD ADC channel

[2]

- **ADC** channel estimates the number of muons in the detector by dividing the total signal charge by the mean charge of a single muon.
- **The charge distribution of one** muon is assumed to follow a lognormal distribution





## ADC Simulation

 $\mathcal{L} = \mathcal{L}$ 

 $\overline{Q}$ 

ADC signals can be described by, [3]

$$
f(t; \mu, \sigma) = \frac{Q_J}{\sum_{i=1}^N \frac{e^{-\frac{(\ln(t_i - t_0) - \mu)^2}{2\sigma^2}} \Delta t}{t_i - t_0}} \cdot \frac{e^{-\frac{(\ln(t_i - t_0) - \mu)^2}{2\sigma^2}} \Delta t}{t_i - t_0}
$$
Single signal

The saturation level of the signal is calculated as,

$$
S_L = N_{\mu} \cdot \frac{Q_J e^{-\mu + \sigma^2/2}}{\sum_{i=1}^N \frac{e^{-\frac{(\ln(t_i - t_0) - \mu)^2}{2\sigma^2}}\Delta t}{t_i - t_0}}
$$

Nμ= 3×362 for the LG channel [2]



## ADC pulse shape



Multiple signals arriving at different times added to get a single pulse

## Injected number of muons at the each channel



## Saturated fraction of events

- **Fiaction of saturated** events increases with energy.
- Detectors saturate less when ADC is included.



#### Current reconstruction method

- **Profile/integrated likelihood method with the detector timing in the counter mode**
- Original likelihood reconstruction method

## Likelihoods used for the hybrid reconstruction method

The MLDF  $\mu(r_i,\vec{p})$  depends on the distance between the detector and the shower axis (r<sub>i</sub>) and the free parameters in  $\vec{p}$ , obtained by maximising the likelihood function.

When binary channel is saturated and the estimated number of muons in the ADC channel is less than 200

$$
\theta_n = \sqrt{\ln\left(1 + \frac{e^{\theta^2} - 1}{n}\right)}
$$
\n
$$
m_n = m + \frac{\theta^2}{2} + \ln\left(\frac{n}{\sqrt{1 + \frac{e^{\theta^2} - 1}{n}}}\right)
$$
\n
$$
L = \sum_{n=1}^{n_{LG}} \frac{1}{\sqrt{2\pi}\theta_n n_{LG}} \cdot \exp\left(\frac{-(\ln(n_{LG}\langle q \rangle) - m_n)^2}{2\theta_n^2}\right) \cdot \frac{e^{-\mu(r_i;\vec{p})}\mu(r_i;\vec{p})^n}{(1 - e^{-\mu(r_i;\vec{p})})n!}
$$

When binary channel is saturated and the estimated number of muons in the ADC channel is between 200 and the LG saturation limit (around 3×362),

$$
\langle n \rangle = \frac{\mu(r_i; \vec{p})}{1 - e^{-\mu(r_i; \vec{p})}}
$$

$$
\sigma^2 = \langle n \rangle \left( (\varepsilon^2 + 1) - \langle n \rangle e^{-\mu(r_i; \vec{p})} \right)
$$

$$
L = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left( -\frac{(n_{LG} - \langle n \rangle)^2}{2\sigma^2} \right)
$$

When the estimated number of muons in the ADC channel is greater than LG saturation limit (around 3×362)

$$
L = 0.5 \times \left(1 - erf\left(\frac{n_{LG} - \langle n \rangle}{\sqrt{2\sigma^2}}\right)\right)
$$

## **Simulations**



- **FICT** For each sampled event, the distance of the detector to the shower axis and arival time of each muon are obtained.
- The average LDF  $\rightarrow$  expected number of muons ( $\mu$ )  $\rightarrow$  actual number of muons
- The LDF is fitted to the detector data by either minimizing the  $x^2$  or by maximizing a likelihood function. In this work, the method of maximizing a likelihood is used.

For each event μ(r) was adjusted using a second Kascade Grande like muon LDF.

$$
\mu = \mu_0 \frac{g(r)}{g(r_0)}
$$

$$
g(r) = \left(\frac{r}{r_1}\right)^{-\alpha} \left(1 + \frac{r}{r_1}\right)^{-\beta} \left(1 + \left(\frac{r}{10r_1}\right)^2\right)^{-\gamma}
$$

$$
\begin{array}{l} \alpha = 0.75 \\ r_1 = 320m \\ \gamma = 2.55 \end{array}
$$

 $r \rightarrow$  distance to the shower axis measured in the shower plane

 $μ<sub>0</sub>$  and β are adjusted by minimizing the function

$$
-2\ln(L_{fit}(\mu_0,\beta)) = -2\sum_i \ln L_i(\mu(r_i,\mu_0,\beta))
$$

#### MLDF fitted to simulated detector data



Iron primaries Time bin 40 ns  $log_{10}(E/eV) = 18.5$  $\theta = 30^\circ$ 

## Examples of events



 $log_{10}(E/eV) = 19$   $log_{10}(E/eV) = 18.75$ 

#### Reconstruction performance **Bias**



#### Standard deviation



#### Coverage



## **Conclusions**

- ADC channel  $\rightarrow$  investigate closer to the shower core.
- $\mathcal{C}_{\mathcal{A}}$ Increases the precision of the UMD.
- Saturation in ADC happens for a very large number of injected muons.
- **Using the new reconstruction method, more events can be reconstructed.**
- The small bias and the low standard deviation achieved allows for a good estimation of μ(450).

## Future work

- Study the profile+ADC method in detail.
- The correlation between  $X_{\text{max}}$  and the total number of muons of the showers can be used to differentiate between a mixed composition scenario from a pure one, in a given energy bin.

$$
r = \frac{cov(X_{max}, N_{\mu})}{\sigma(X_{max})\sigma(N_{\mu})}
$$

 $\blacksquare$  N<sub>u</sub> will be calculated from the hybrid reconstruction method explained in the previous slides.

#### References

[1] Calibration of the underground muon detector of the Pierre Auger Observatory, The Pierre Auger collaboration (2021)

[2] Design, upgrade and characterization of the silicon photomultiplier front-end for the AMIGA detector at the Pierre Auger Observatory, The Pierre Auger collaboration (2021)

[3] A.M.Botti, GAP2020\_003

## Backup slides

## AMIGA UMD



[1]

- **UMD** measures the fall of muon density with the distance to the shower axis  $\rightarrow$  the lateral distribution function (LDF).
- **Measures showers between 10<sup>16.5</sup> eV to 10<sup>19</sup>** eV and events up to 45° zenith angle.
- The detectors are buried 2.3 m underground
- **Counter mode measures low densities and** number of muons is determined by counting signals above a threshold.
- **The integrator mode measures high muon** densities.

#### Detector simulation





[3]

#### Simulation of the detectors



## Muon time distribution at different shower axis distances



We use 40 ns time window for the analysis 12 inhibition windows  $\times$  3.125 ns time = 37.5 ns

40 ns time window For larger  $r \rightarrow$  time distribution becomes wider



Comparison of log-normal and Gaussian distributions in 20 region

