# Extraction of Muons and Validation of MC

# for the First G-APD Cherenkov Telescope

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SFB 876 Providing Information by Resource-Constrained Data Analysis



CTA Calibration Meeting 2015

**Extraction of Muons** 

**Training Data** 

CORSIKA MC

Feature Extraction with FACT-Tools

**Machine Learning** 

Validation of MC

**Time Resolution** 

PSF

Conclusion



[Miguel Claro]





- → Supervised models need training data with known truth:
  - → Particle type (classification)
  - → Energy (regression)

- → This can be hard for muon ring images
  - → Hadronic primary → muon ring image
  - $\rightarrow$  Muon primary  $\rightarrow$  decay or air shower

# $\Rightarrow$ Signal in background MC and background in signal MC

To get a clean muon ring sample, we use the following settings:

Energy range10 GeV to 1 TeVSpectral index-2.7Starting altitude300 m above FACTZenith distance0°Maximum Impact Parameter1.75 m (mirror radius)ViewCone2°



#### Simulation with CORSIKA - Protons

Energy range 100 GeV to 200 TeVSpectral index -2.7

**Zenith distance**  $0^{\circ} - 30^{\circ}$ 

**ViewCone**  $5^{\circ}$  (mirror radius) Discard events if more than 50% of the Cherenkov photons were emitted by muons.



- → Joined forces with the Artificial Intelligence Group / Department for Computer Sciences
- → Extension of the streams-framework developed in Dortmund
- → Easy deployment (single .jar-file)
- → Rapid prototyping with Event-Viewer
- → Process is designed with xml

#### **Current** Calibration and feature extraction

#### **Planned** Add application of trained model, proof of concept in:

Christian Bockermann et al. "Online Analysis of High-Volume Data Streams in Astroparticle Physics". In: *Machine Learning: ECML 2015, Industrial Track.* Springer Berlin Heidelberg, 2015

#### sfb876.tu-dortmund.de/FACT/

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#### A Possible Data Stream



- → Standard Hillas parameters also helpful for muon identification
- → Timing features very discriminative
- → Dedicated muon features improve the performance



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- → Shape recognition algorithm
- → Transformation from image space into discrete parameter space

 $Photons(pixel) \rightarrow Photons(R, x_0, y_0)$  (1)

- → Pixels "vote" for rings
- → Strong features can be calculated from shape of parameter space
- → Computationally expensive

#### Hough Transform: Muon Ring Image



Extraction of Muons and Validation of MC – Feature Extraction with FACT-Tools

#### Hough Transform: Proton Shower



Extraction of Muons and Validation of MC – Feature Extraction with FACT-Tools

#### Hough Transform: Features



#### Machine Learning: Decision Trees

#### Training

- → Needs labeled data
- → Pick the most discriminative feature
- → Split at the best possible cut
- → Repeat until only pure data sets are left or a given maximum depth is reached

#### Application

- → Unknown event goes through the tree
- → Leaf determines classification



#### Machine Learning: Random Forests

#### **Problems with single DTs**

- → Tend to overtraining
- → Rather large reaction on small fluctuations in training data

#### Solution

- → Train many trees on slightly different training data (*Bagging*)
- → Consider only a random subset of features at each node
- → Confidence: proportion of the trees that voted for target class



#### **Performance Evaluation**



#### Precision

 $\texttt{tp} \,/\, (\texttt{tp} + \texttt{fp})$ 

→ also known as purity

#### Recall

 $\texttt{tp} \: / \: (\texttt{tp} + \texttt{fn})$ 

→ also known as efficiency or true positive rate

 $\label{eq:False positive rate} \begin{array}{l} \mbox{False positive rate} \\ \mbox{fp} \ / \ (\mbox{tn} + \mbox{fp}) \end{array}$ 

## Choosing a Confidence Cut



- → All events with confidence > cut are classified as signal
- → Classification is trade-off between precision and recall
- → Which is more important for your task?

# Choosing a Confidence Cut



Where do the uncertainties come from?

- → All events with confidence > cut are classified as signal
- → Classification is trade-off between precision and recall
- → Which is more important for your task?

- → Performance values without uncertainties are meaningless
- → Split training data in N subsamples
- → Train on N − 1, evaluate performance on the other part
- → Calculate mean and standard deviation of performance values
- → Large uncertainties implicate overfitting



- → Extract Muons from telescope data and proton Monte-Carlo
- → Compare attribute distributions
- → Muons especially interesting for
  - $\rightarrow$  Time resolution
  - $\rightarrow$  PDE
  - $\rightarrow$  PSF

# Timing

- → Standard deviation of arrival times indicates time resolution
- → Dependent on reconstruction and NSB levels
- → Rather huge mismatch at FACT in the beginning



# First ideaFixed offsets between<br/>channelsMeasurementLED-Lightpulser<br/>Calibration runsResultBetter time resolution on<br/>data, but not close to MC



# Timing

#### First idea Fixed offsets between channels Measurement LED-Lightpulser Calibration runs

Result Better time resolution on data, but not close to MC



# Timing



- → Possible explanations:
  - → G-APD time jitter
  - → Scattering in the Winston cones
  - → Electronics
  - → Residual of DRS time calibration
  - $\rightarrow \dots$



#### **Point-Spread-Function**



- → Gaussian fit to radial light distribution
- $\rightarrow \sigma$  of the gaussian indicates PSF



Sebastian Müller et al. "FACT – Novel mirror alignment using Bokeh and enhancement of the VERITAS SCCAN alignment method". In: *Proceedings of the 34th ICRC*. 976. 2015

- → Machine Learning methods well suited for muon extraction
- → streams-framework allows online application of such methods
- → Pipe out muons for calibration from the live data stream
- → Important cross check for MC