

# Machine Learning in Gamma-Ray Astronomy

More than just Background Suppression

Maximilian Nöthe

TU Dortmund

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Introduction to FACT

Data Preparation

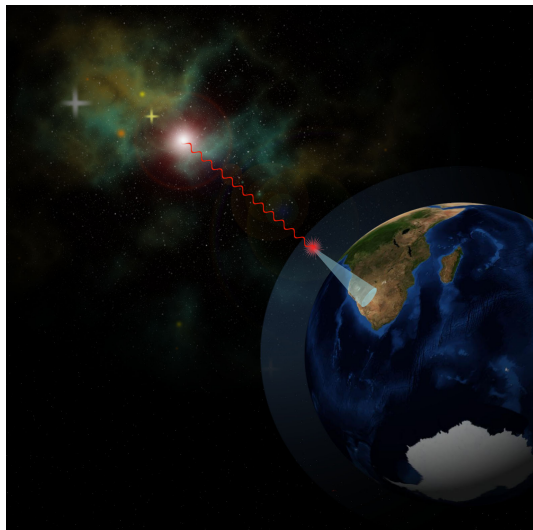
Current use of machine learning methods

Can we use Deep Learning?

Conclusion

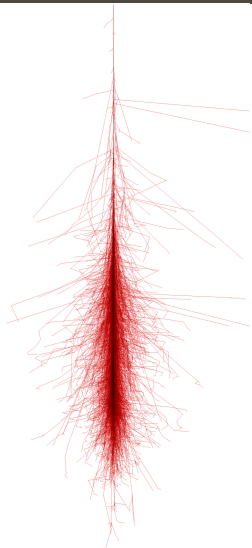
# Imaging Air Cherenkov Telescopes

- Gamma-rays and hadrons produce extensive air showers
- Telescopes detect the Cherenkov light on the ground
- Usually 1000–10000 hadrons per gamma-ray
- Reconstruction tasks:
  - Particle type ( $\gamma$  / hadron)
  - Energy
  - Origin

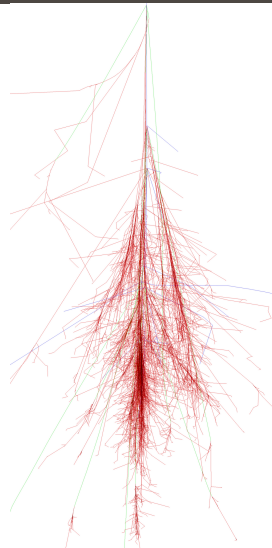


[[www-ucjf.troja.mff.cuni.cz/~hess/img/sprska4.jpg](http://www-ucjf.troja.mff.cuni.cz/~hess/img/sprska4.jpg)]

# Feature extraction



100 GeV Gamma Shower

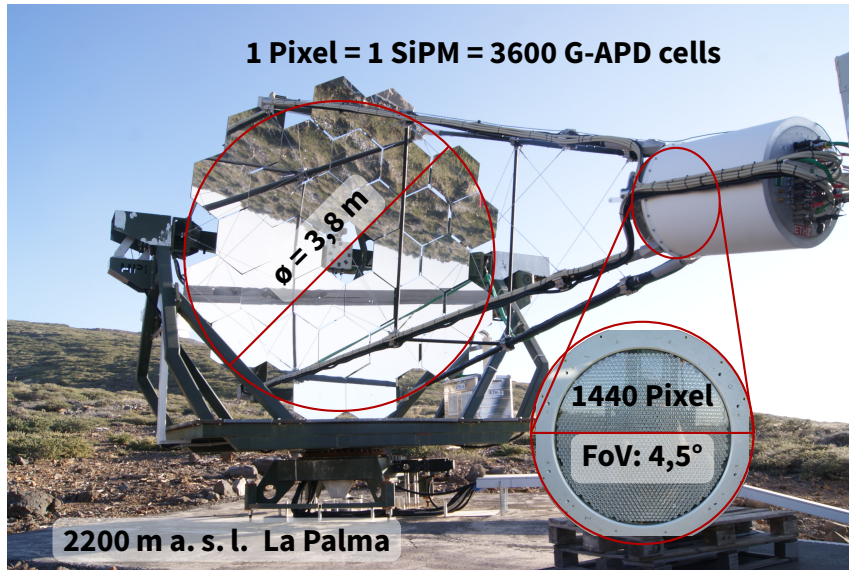


100 GeV Proton Shower

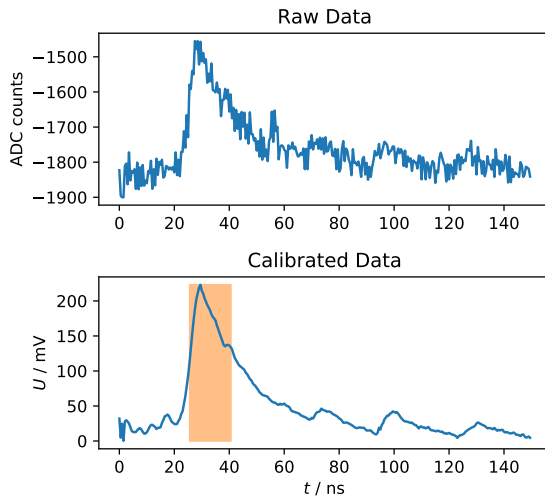
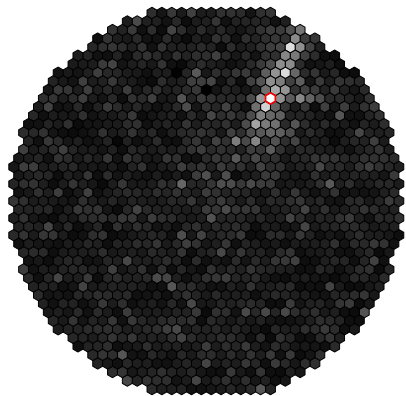
[from: [www.ikp.kit.edu/corsika](http://www.ikp.kit.edu/corsika)]



# FACT – The First G-APD Cherenkov Telescope

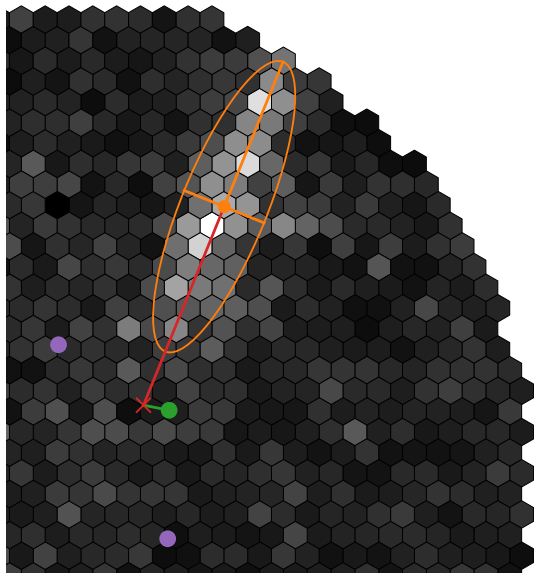


# Raw Data



Either measured telescope data or CORSIKA plus detector simulation.

# Feature extraction



- Further data reduction from photons and arrival times to features
- Classical Hillas features describing the light distribution
- Descriptive statistics of the light and time distributions
- 40–60 features in the end
- Source dependent features are calculated also for “off” positions

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- We are working on using machine learning methods
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[portal.cta-observatory.org/PublishingImages/Sites/Spain]

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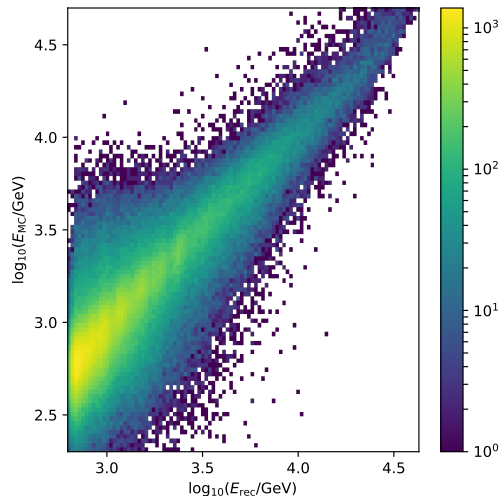
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[[www.cta-observatory.org/about/array-locations/la-palma](http://www.cta-observatory.org/about/array-locations/la-palma)]

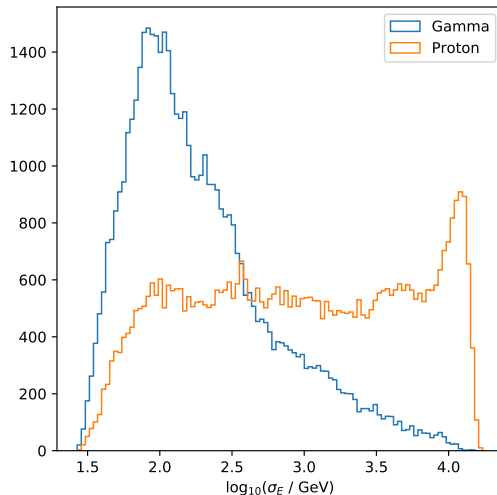
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- We use `scikit-learn`'s `RandomForestRegressor`
- Trained on gamma-ray Monte Carlo data
- Use Standard Deviation of tree answers for Separation



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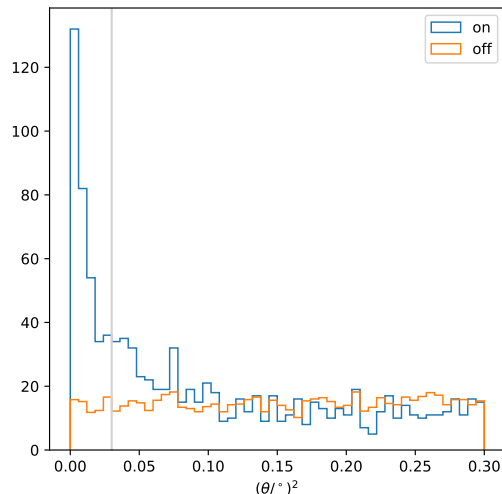
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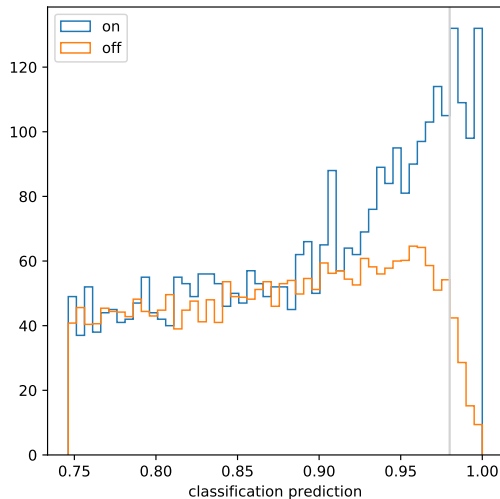
# Particle Classification – Classical Approach

- Leave out any features dependent on source position
- Target class: gamma-rays
- Source detection with  $\theta^2$  using on/off data
- RandomForestClassifier



# Particle Classification – New Approach

- Target class: gamma-rays from the source
- Source detection with classification output
- Requires application of the model for each off position
- only possible for known point sources



# Online application of machine learning models

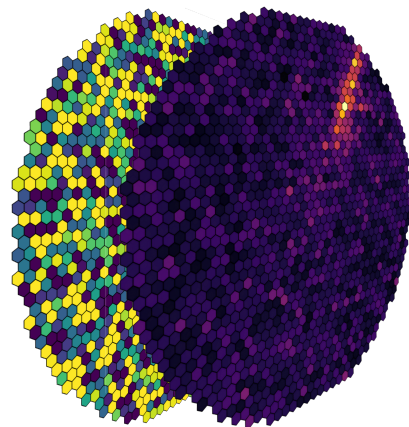
- Online application of models trained with scikit-learn in our java analysis framework
- Export models to PMML
- Apply models in java

<https://github.com/fact-project>

**SFB 876** Providing  
Information by Resource-  
Constrained Data Analysis

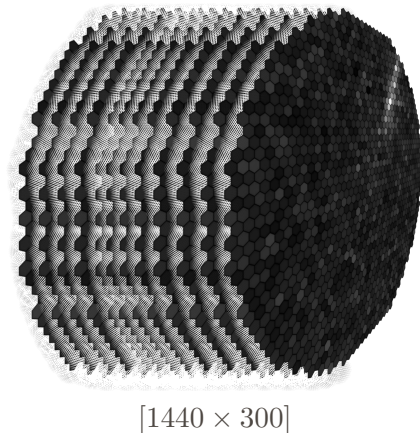


- Improve on our separation & regression tasks
- Reduce runtime for live analysis → Alert other experiments
- Directly go from Photons and Arrival Times to target variables
- Maybe even directly from the raw data?
- Current problems:
  - No support for hexagonal grids
  - How to deal with hexgrids?



$[1440 \times 2]$

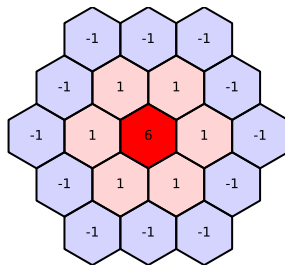
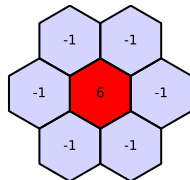
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# Difficulties with hexgrids

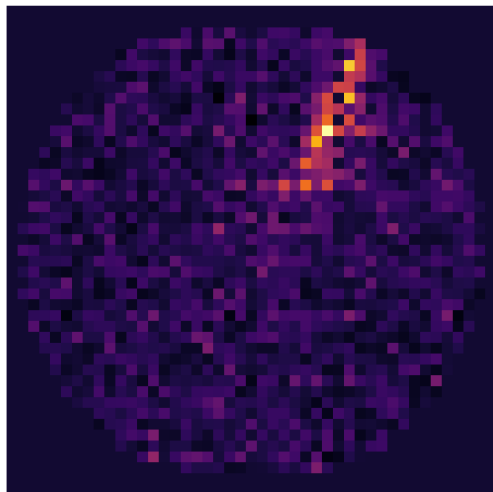
- Borders of the camera not regular
- Going from  $\text{height} \times \text{width}$  to radius for the filter
- Which coordinate system to use?
- Can we contribute hexgrid support to keras / tensorflow?

If you want to learn more about hexgrids:  
[www.redblobgames.com/grids/hexagons/](http://www.redblobgames.com/grids/hexagons/)



# Square pixel image approximation

- We want to start first experiments with square pixel approximations
- Not taking full advantage of the geometry
- Probably do not gain much runtime performance



- Classical machine learning approaches work very well
- We are working on expanding the use of ML models to more reconstruction tasks
- Constantly improve our shallow learning methods
- We just started to look into Deep Learning