





Nested Machine Learning Models for CTA

Master Thesis Half-Time Talk

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December 16, 2022

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Overview

Introduction

Scaled Parameters and Feature Selection

Energy Regression

Gamma-Hadron Classification

Origin Reconstruction

Performance

Outlook



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Introduction



Imaging Atmospheric Cherenkov Telescopes (IACTs)





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Imaging Atmospheric Cherenkov Telescopes (IACTs)

y-ray entering the atmosphere





Imaging Atmospheric Cherenkov Telescopes (IACTs)





Imaging Atmospheric Cherenkov Telescopes (IACTs)









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The Cherenkov Telescope Array

- CTA South (Paranal Observatory, Chile)
- CTA North (La Palma)
 - 4 LSTs + 9 MSTs
- Large-Sized Telescope (LST)
 - 23 m mirror diameter
 - 4.3° FoV
- Medium-Sized Telescope (MST)
 - 11.5 m mirror diameter
 - 7.7° FoV



[CTAO]



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Dataprocessing using ctapipe R1 - Pixel 314 Pixel 42 Pixel 314 Data Volume Reduction DL0 Pulse Pixel 42 Pixel 314 Extraction R0 # Photons Time / ns Calibration - 35 dec time event energy gammaness ra - 30 1500 0.82 83.6 22.1 59024.63123 - 25 G - 20 2 400 0.73 83.5 21.9 59024.64183 - 15 5 680 0.92 83.7 22.0 59024.67093 -10 -5 0 Reconstruction DL2 Image Cleaning (DL1a) event width length ... intensity Time / ns # Photons 0.15 0.52 1253.1 0 0 05 0.12 ... 321.3 2 - 35 - 30 5 0.08 0.19 ... 512.7 15 - 25 - 20 DL1b - 15 Parametrization - 10 -5

[A. Knierim, M. Linhoff]

Introduction



The disp Method

- Monoscopic origin reconstruction for IACTs
- Assume main shower axis to be correctly reconstructed →source position on axis
- Train regressor to estimate distance from image center of gravity (|disp|/norm)
- Train classifier to decide between remaining two possibilities (sign)



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Scaled Parameters and Feature Selection



Scaled parameters

Compare image length (l) and width (w) with expectation value and variance from simulations as function of image charge (q), impact distance (d), and telescope type (t):

$$SL = \frac{l(q, d, t) - \langle l(q, d, t) \rangle}{\sigma_l(q, d, t)}$$

$$SW = \frac{w(q, d, t) - \langle w(q, d, t) \rangle}{\sigma_w(q, d, t)}$$
Easy combination for stereo observations:

$$MSL = \frac{\sum_{\text{tels}} SL}{\sqrt{n_{\text{tels}}}}$$

$$MSW = \frac{\sum_{\text{tels}} SW}{\sqrt{n_{\text{tels}}}}$$

$$MSW = \frac{\sum_{\text{tels}} SW}{\sqrt{n_{\text{tels}}}}$$



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MRMR feature selection – Energy

At each iteration i compute $score_i$ for all f not yet selected

 $score_i(f) = \frac{relevance(f | target)}{redundancy(f | f_{already selected})}$

and add the feature with the highest score.

Energy features (15)

- peak_time_std
- intensity_std
- timing_deviation
- hillas_length
- concentration_pixel
- leakage_intensity_width_2
- log_abs_timing_slope
- area
- concentration_cog
- log_tel_impact_distance
- hillas_width
- log_intensity
- morphology_n_pixels
- HillasReconstructor_h_max
- leakage_pixels_width_2





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MRMR feature selection – Particle Id

At each iteration i compute $score_i$ for all f not yet selected

 $score_i(f) = \frac{relevance(f | target)}{redundancy(f | f_{already selected})}$

and add the feature with the highest score.

Particle Id features (15)

- scaled_width
- HillasReconstructor_h_max
- log_tel_impact_distance
- hillas_width
- morphology_n_islands
- timing_deviation
- area
- scaled_length
- RandomForestRegressor_energy
- log_RandomForestRegressor_energy
- leakage_pixels_width_2
- hillas_length
- peak_time_kurtosis
- RandomForestRegressor_tel_energy
- morphology_n_pixels

scaled width -				363938.68
hillas width -			267309.59	
timing deviation -		192426.30		
morphology n islands -		162477.95		
area		145176.59		
scaled length -	49410.90			
hillas length -	32163.32			
RandomForestRegressor energy -	27238.81			
log RandomForestRegressor energy -	25832.13			
RandomForestRegressor tel energy -	24490.29			
- elexia n vpolodarom	22912.03			
HillasReconstructor h max-	21919.30			
leakage pixels width 2 -	20700.08			
concentration cog	17902.32			
log tel impact distance -	15403.86			
log abs timing slope -	14666.30			
intensity kurtosis -	7138.11			
leakage intensity width 2 -	6386.02			
peak time kurtosis -	5901.59			
peak time std -	5836.81			
concentration core	5818.17			
log intensity -	5464.32			
HillasReconstructor average intensity -	4437.07			
intensity skewness -	3933.72			
log RandomForestRegressor tel energy -	3912.96			
hillas intensity -	1819.92			
intensity std -	1441.63			
intensity mean -	1015.68			
hillas kūrtosis -	262.44			
hillas r-	245.68			
intensity max-	81.78			
concentration pixel	66.22			
peak time skewness -	48.63			
HillasReconstructor tel impact distance	14.35			
timing slope -	2.23			
hillas skewness	0.14			
-				
	1			

0 50000 100000 150000 200000 250000 300000 350000 400000 *F*-test score



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disp features (19)

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- log_tel_impact_distance
- log_RandomForestRegressor_tel_energy
- log_abs_timing_slope
- peak_time_std
- concentration_pixel
- hillas_length
- concentration_cog
- timing_deviation
- scaled_length
- HillasReconstructor_h_max
- RandomForestRegressor_energy
- area
- peak_time_kurtosis
- RandomForestRegressor_tel_energy
- timing_slope
- hillas_skewness
- HillasReconstructor_core_x
- HillasReconstructor_core_y





Scaled Parameters and Feature Selection



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sign classifier



Scaled Parameters and Feature Selection



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Concept







Energy Regression



Energy Regression (Telescope)

Randomized hyperparameter optimization yields:



5-fold cross-validation:

551 635 LST events

1 199 782 MST events





Energy Regression (Telescope)

Randomized hyperparameter optimization yields:



5-fold cross-validation:

551 635 LST events

1 199 782 MST events

Energy confusion for MST $(R^2 = 0.8641 \pm 0.0044)$





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Feature Importance for LST

0.2

0.4

feature importance

0.6

peak_time_std -

0.0

Energy Regression (Telescope)

Randomized hyperparameter optimization yields:



1 199 782 MST events

- C. C.



Energy Regression (Telescope)

Randomized hyperparameter optimization yields:



log_target: True

5-fold cross-validation:

551 635 LST events

1 199 782 MST events



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Configuration array energy regressor

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Energy Regression (Array)

- Use (averaged) telescope predictions and array-wide features
- No hyperparameter optimization (yet)
- 5-fold cross-validation on 461 969 events

```
model cls: RandomForestRegressor
model_config:
    n estimators: 200
    max features: "sqrt"
    min_samples_leaf: 0.00001
   n jobs: 40
log_target: True
features:
    - n_telescopes_HillasReconstructor
    - n LST HillasReconstructor
    - n_MST_HillasReconstructor
    - mean scaled length
    - mean scaled width
    - HillasReconstructor_core_x
    - HillasReconstructor core y
    - HillasReconstructor_average_intensity
    - HillasReconstructor h max
    - HillasReconstructor alt
    - HillasReconstructor_az
    - RandomForestClassifier prediction
    - RandomForestRegressor energy
```



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- RandomForestClassifier_prediction
- RandomForestRegressor_energy



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```





Gamma-Hadron Classification



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Gamma-Hadron Classification (Telescope)

Randomized hyperparameter optimization yields:

Configuration particle classifier	
<pre>ParticleClassifier: model_cls: RandomForestClassifier model_config: n_estimators: 69 max_features: 0.5227 max_samples: 0.7138 min_samples_leaf: 0.000013 n_jobs: 40</pre>	

- LST: 552 754 signal + 561 171 background events
- MST: 1199267 signal + 1122374 background events





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Configuration array particle classifier

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Gamma-Hadron Classification (Array)

- Use (averaged) telescope predictions and array-wide features
- No hyperparameter optimization (yet)
- 5-fold cross-validation on 910 387 events

```
model cls: RandomForestClassifier
model_config:
 n estimators: 200
 max features: "sqrt"
 min samples_leaf: 0.00001
 n jobs: 40
features:
  - n telescopes HillasReconstructor
  - n_LST_HillasReconstructor
  - n MST HillasReconstructor
  - mean scaled length
  - mean_scaled_width
  - HillasReconstructor core x
  - HillasReconstructor_core_y
  - HillasReconstructor_average_intensity
  - HillasReconstructor h max
  - HillasReconstructor alt
  - HillasReconstructor az
   RandomForestClassifier prediction
  - RandomForestRegressor_energy
```



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   RandomForestClassifier prediction
   RandomForestRegressor_energy
```





Origin Reconstruction



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Origin Reconstruction using disp

Randomized hyperparameter optimization yields:



- 552 754 LST events
- nts 🛛 🛯 1 199 267 MST events



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Origin Reconstruction using disp

Randomized hyperparameter optimization yields:



 $E_{\rm true} / {\rm TeV}$

- 552 754 | ST events 1199 267 MST events



Randomized hyperparameter optimization yields:

Configuration disp reconstructor

```
norm_cls: RandomForestRegressor
norm_config:
    n_estimators: 69
    max_features: 0.5227
    max_samples: 0.7138
    min_samples_leaf: 0.000013
    n_jobs: 40
```

```
log_target: True
```

```
sign_cls: RandomForestClassifier
sign_config:
    n_estimators: 343
    max_features: 0.6587
    max_samples: 0.5815
    min_samples_leaf: 0.000035
    n_jobs: 40
```

5-fold cross-validation:

552 754 LST events

ents 🛛 🗧 1 199 267 MST events



feature importance



Randomized hyperparameter optimization yields:



feature importance

5-fold cross-validation:

552 754 | ST events 1199 267 MST events



Randomized hyperparameter optimization yields:

Configuration disp reconstructor

```
norm cls: RandomForestRegressor
norm_config:
 n estimators: 69
 max features: 0.5227
 max samples: 0.7138
 min samples leaf: 0.000013
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552 754 | ST events 1199 267 MST events



feature importance

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 n jobs: 40
```

Feature Importance for sign for MST (accuracy = 0.9395 ± 0.0005)



feature importance

5-fold cross-validation:

552 754 | ST events 1199 267 MST events



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Origin Reconstruction (Array)

- Predict 3D cartesian position based on unit-sphere (1, alt, az)
- Use (averaged) telescope predictions and array-wide features
- No hyperparameter optimization (yet)
- 5-fold cross-validation on 461 212 events

Configuration (every) array origin regressor

```
model cls: RandomForestRegressor
model config:
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 max features: "sqrt"
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 n jobs: 40
features:
  - n telescopes HillasReconstructor
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  - mean scaled length
  - mean_scaled_width
  - HillasReconstructor core x
  - HillasReconstructor core y
  - HillasReconstructor_average_intensity
  - HillasReconstructor h max
  - HillasReconstructor alt
  - HillasReconstructor_az
   disp alt
  - disp az
  - disp_ang_distance_uncert
```

- RandomForestClassifier_prediction
- RandomForestRegressor_energy

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Origin Reconstruction (Array)

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```

- HillasReconstructor_h_max
- HillasReconstructor_alt
- HillasReconstructor_az
- disp_alt

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- disp_az
- disp_ang_distance_uncert
- RandomForestClassifier_prediction
- RandomForestRegressor_energy



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Performance



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Gamma-Hadron Performance (Energy-Dependent)



Array classifier



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Gamma-Hadron Scores





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Gammaness and θ Cuts





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Gammaness and θ Cuts







Energy Migration





Energy – Bias and Resolution



Gammaness cut (40% eff.) & Θ cut (68% eff.) & FoV offset < 1.0 deg



Sensitivity

⇒ This and all following plots use gammaness cuts optimized for maximum sensitivity!





Angular Resolution



Optimized gammaness cut



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Effective Area







Outlook





TODO

- Separate dataset for array models → optimize hyperparameters
- Try other methods for averaging disp predictions (→ Lukas' master thesis)
- Telescope models using only mono features → include "mono" events
- Try other ML algorithms (e.g. boosted decision trees) incl. hyperparameter optimization









Backup









Timing Parameters vs Impact Distance







Quality Cuts





Background Rejection





Angular Resolution – More Plots



Α



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Angular Resolution – 4x4

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Error of Mean disp Predicitons

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Outlook