

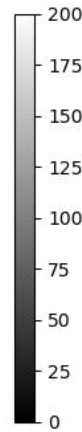
# BGRem: Background noise removal for astronomical images

## Test Data

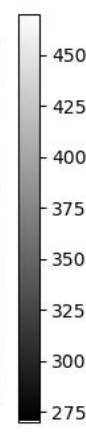
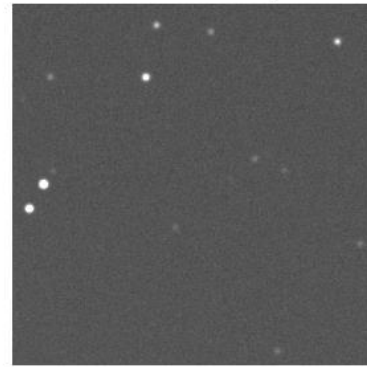
Predicted no background



simulated no background



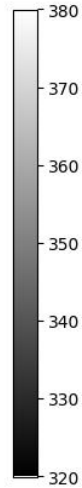
simulated with background



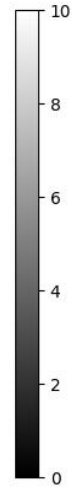
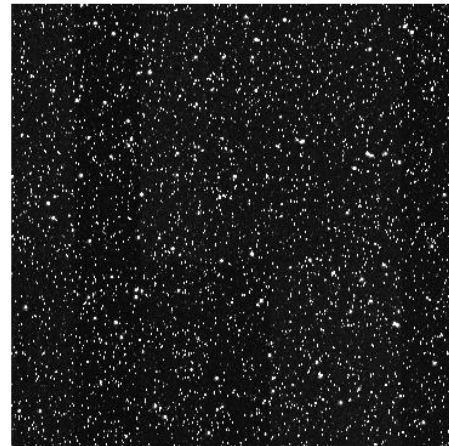
- Diffusion model with attention U-net
- Trained on simulated MeerLICHT images
- Can remove complex backgrounds

## Real images

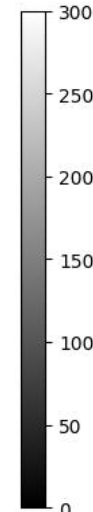
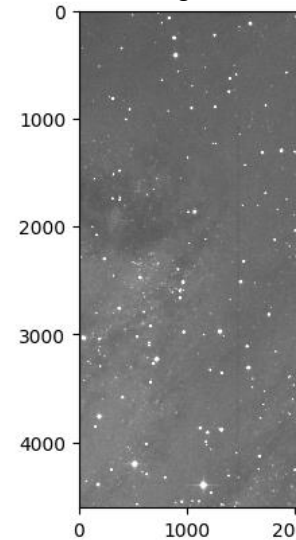
Original



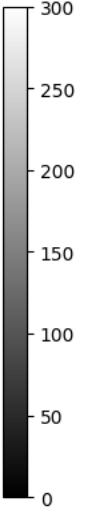
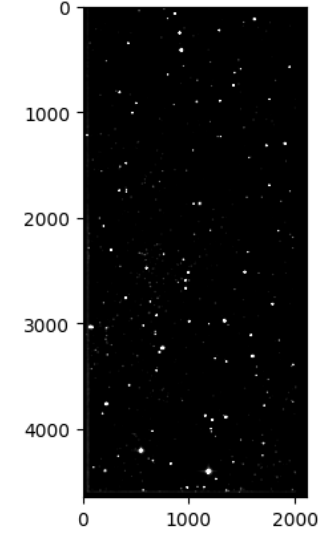
Background removed



Original

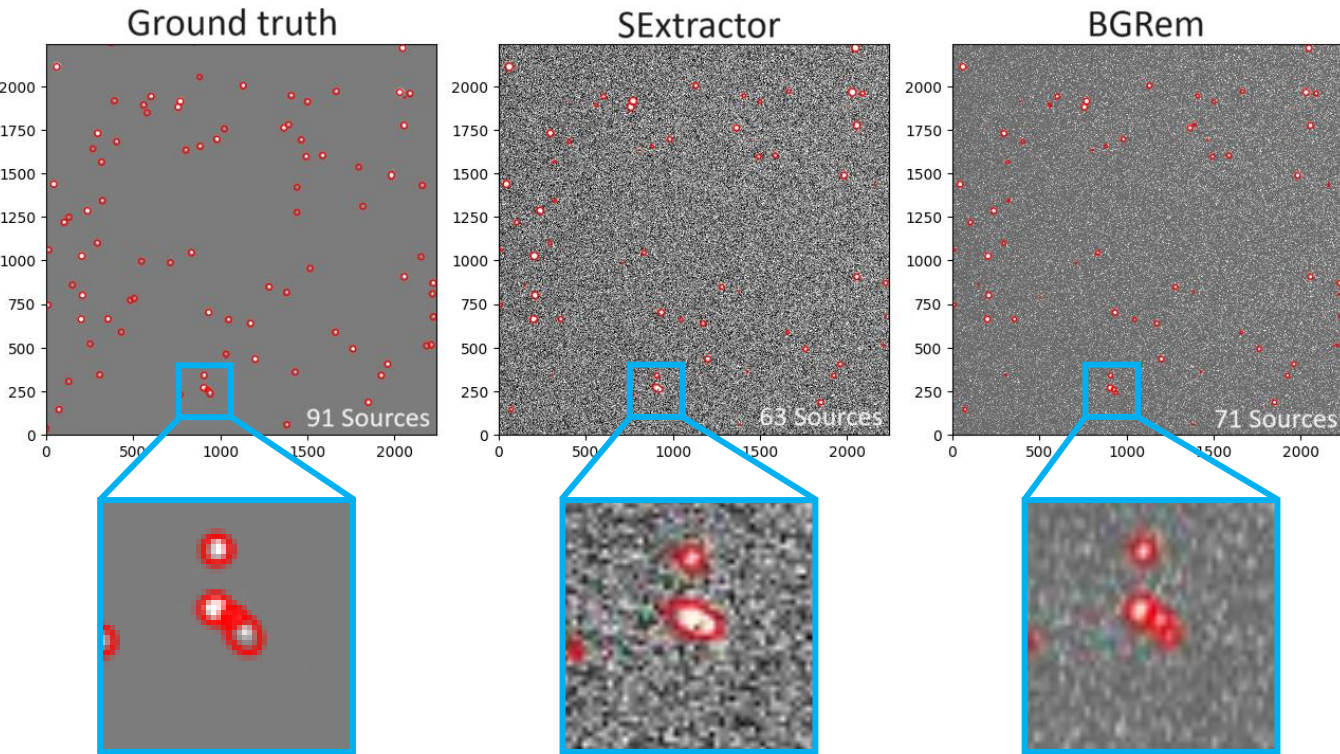


Background removed

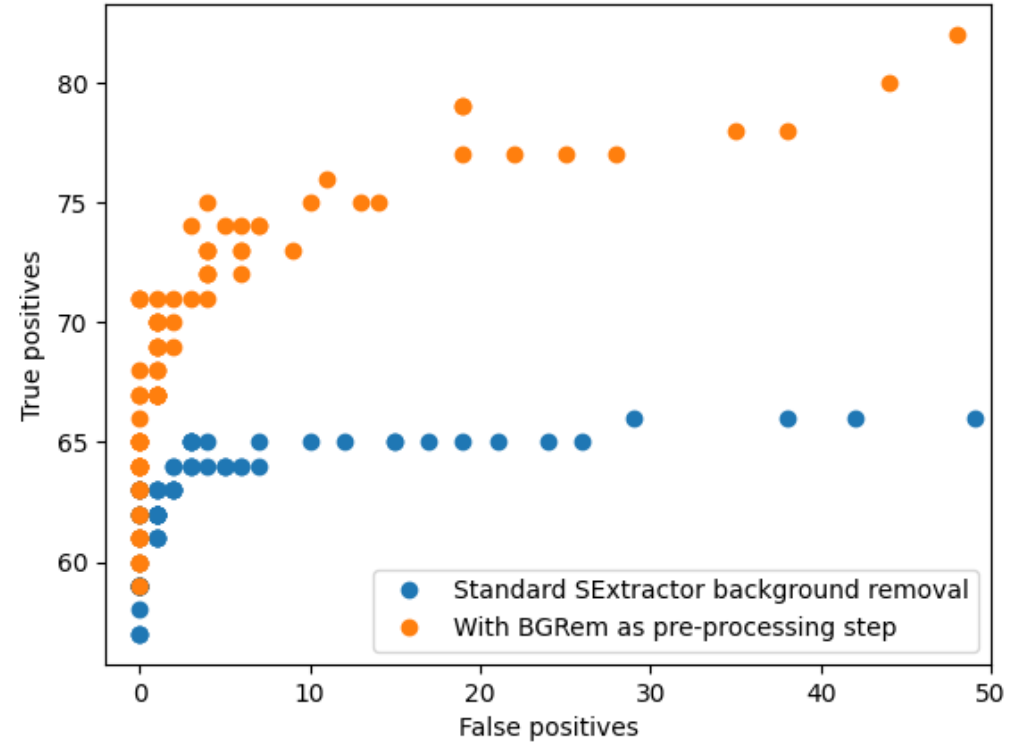


# Source Localisation

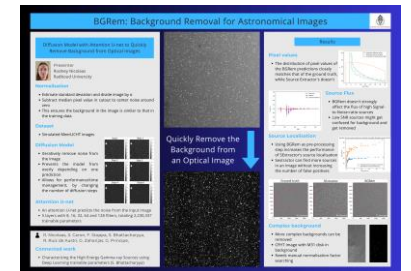
- Find more sources with SExtractor
- No increase in bogus sources
- Helps distinguish sources with close proximity



Number of true sources found as a function of number of bogus sources found

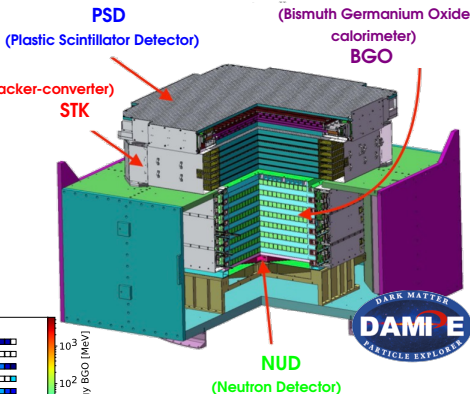


Poster Thursday lunch  
Location 88

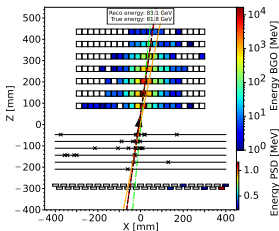


## MOTIVATION

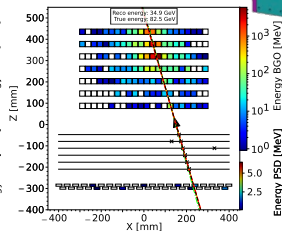
- **Goal:** Obtain an efficient  $\gamma$ -ray event selection
- Most abundant cosmic-ray component: **protons**
- Main difference between  $\gamma$ -rays and protons: **shower topology** in the BGO calorimeter



MC Gamma



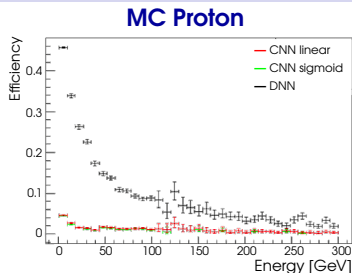
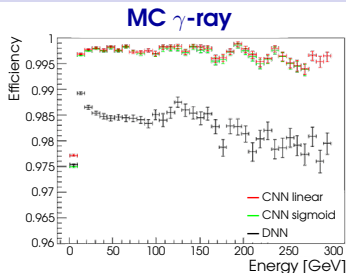
MC Proton



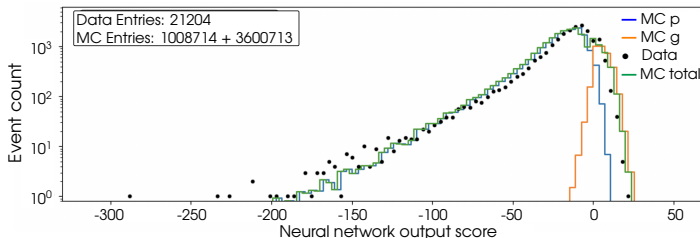
→ ideal ML tool to recognise features from images:

⇒ Convolutional Neural Network!

## CNN MODEL, TRAINING, SCORE



- This method significantly outperforms all the existing algorithms, both in  $\gamma$ -ray efficiency and proton rejection! **Output:** CNN score  $[-\infty, +\infty]$  validation



*Want to learn more details? Please visit my poster in session B :)*