

Early use of ML for NR/ER discrimination from raw pixel data

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Intro

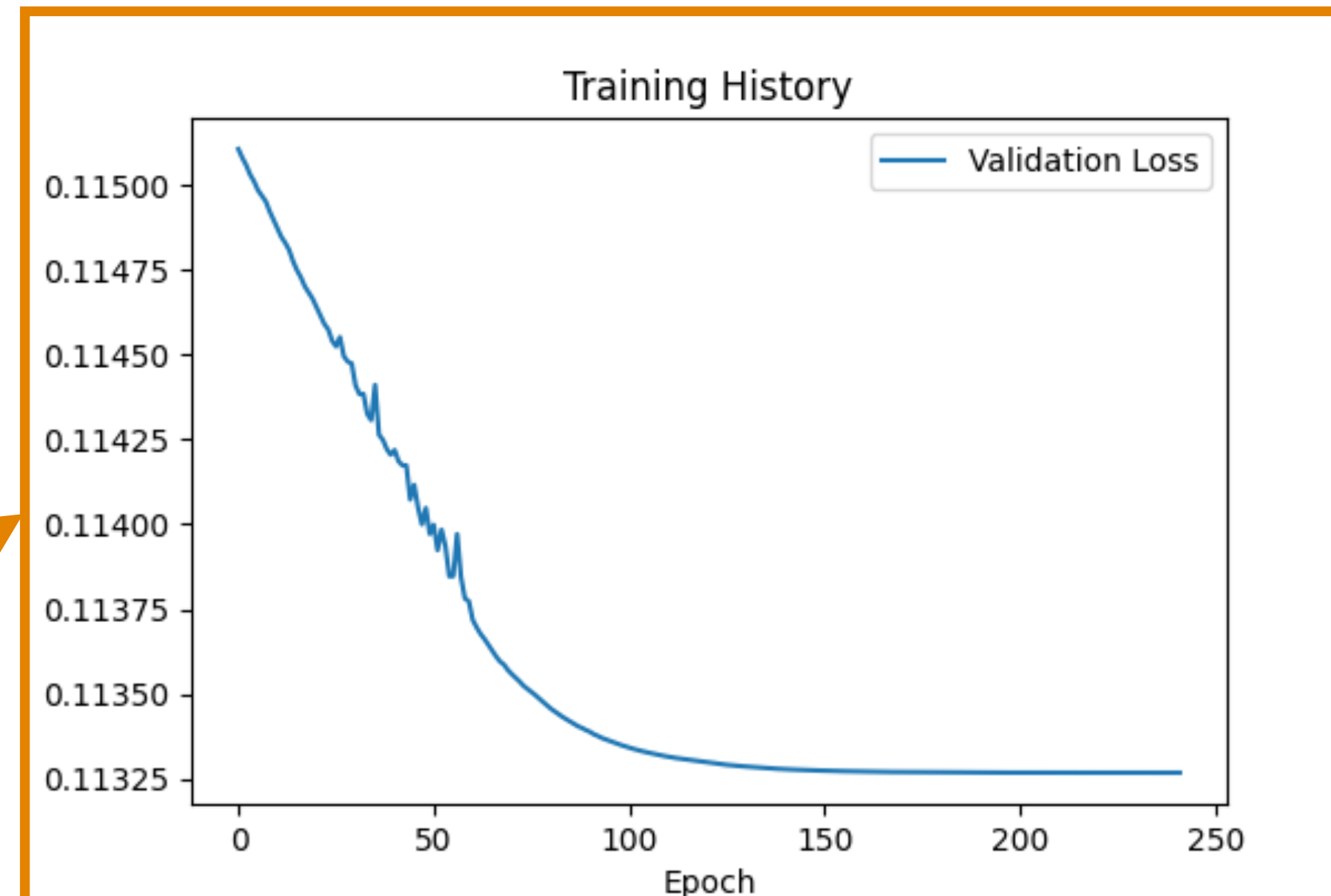
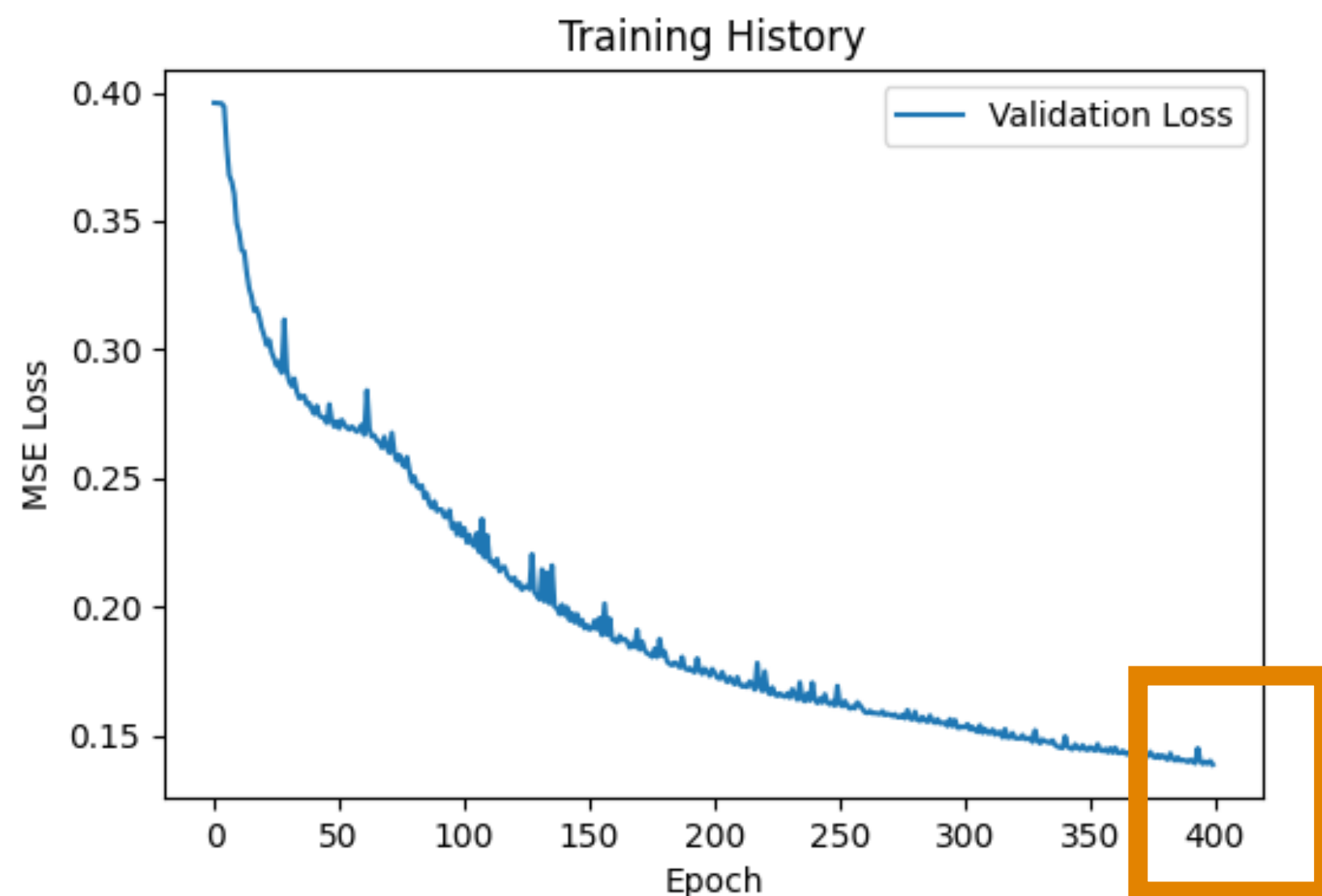
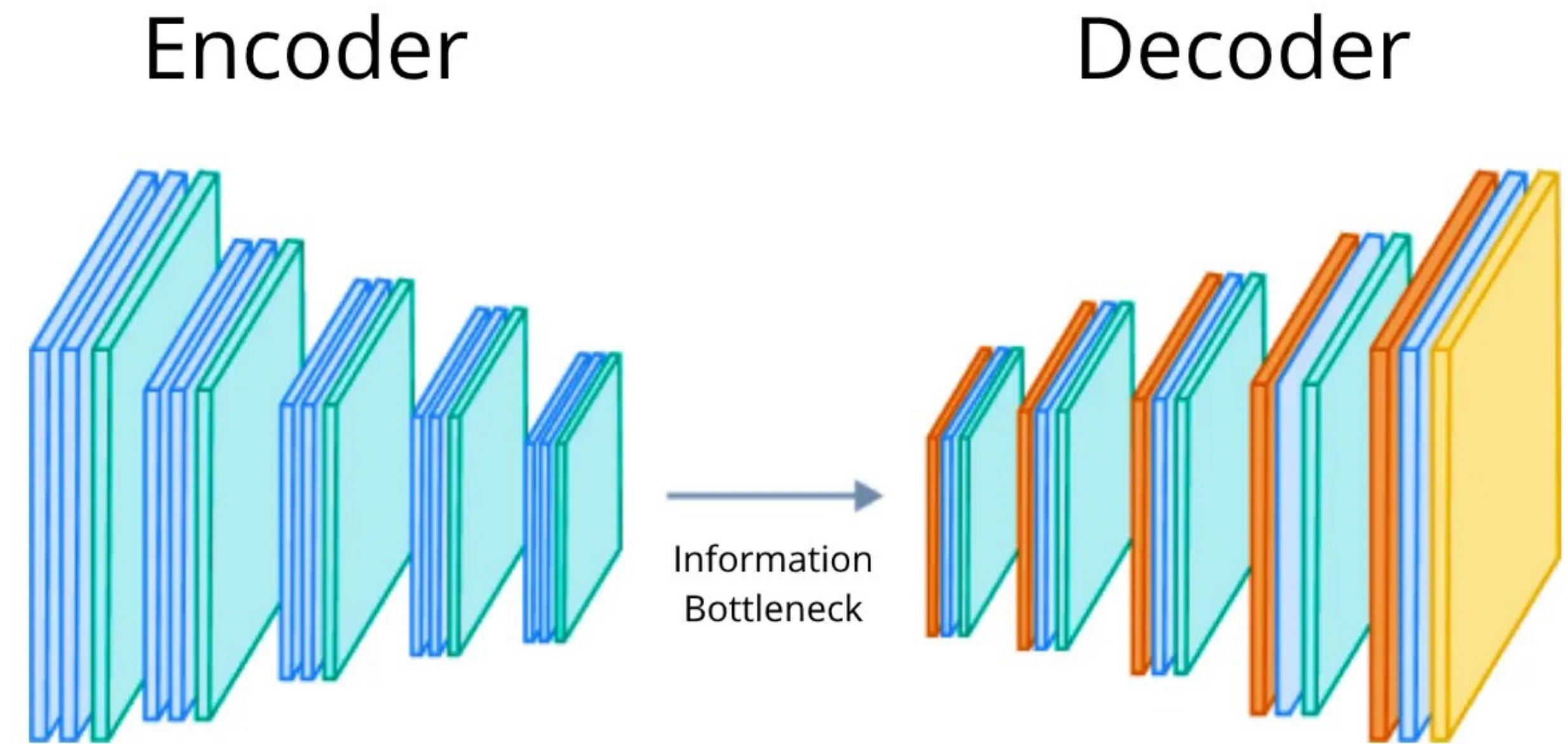
Why ML

- We have lots of data at our disposal (unlabelled)
- Goal: Capturing (and later discriminating) different kinds of events
- Advantages: fast inference time, possibility to fine tune on different configurations of the detector
- Possibility to add montecarlo events to the dataset to increase performance



The Idea

- Train an autoencoder on pedestal images (at the moment: 1 pedestal run)
- In test mode: send tracks with pictures
→ higher reconstruction error
→ anomaly



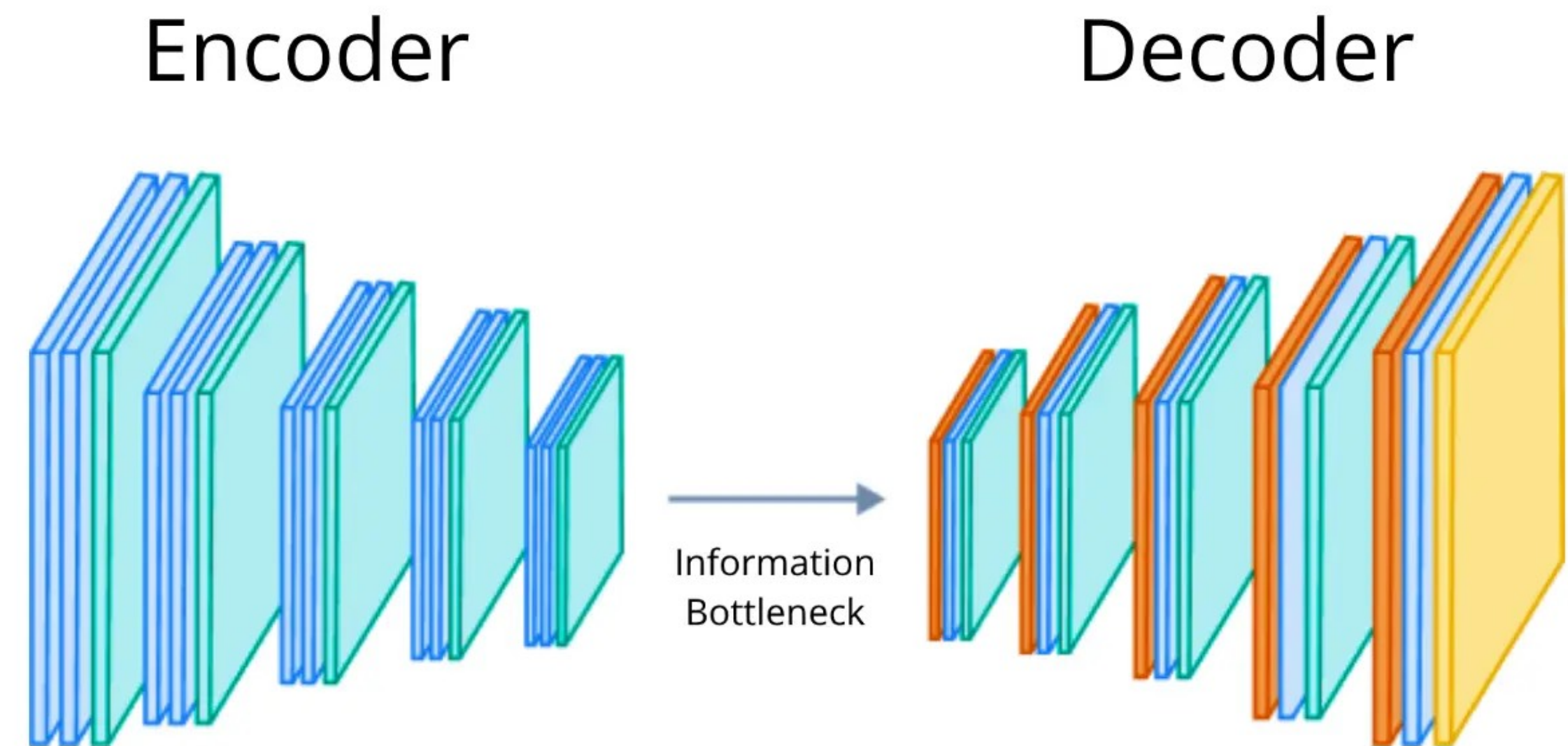
Train Time:
O(5hrs)

Inference time:
O(20ms)/image

GPU used:
M1 Pro 16Gb

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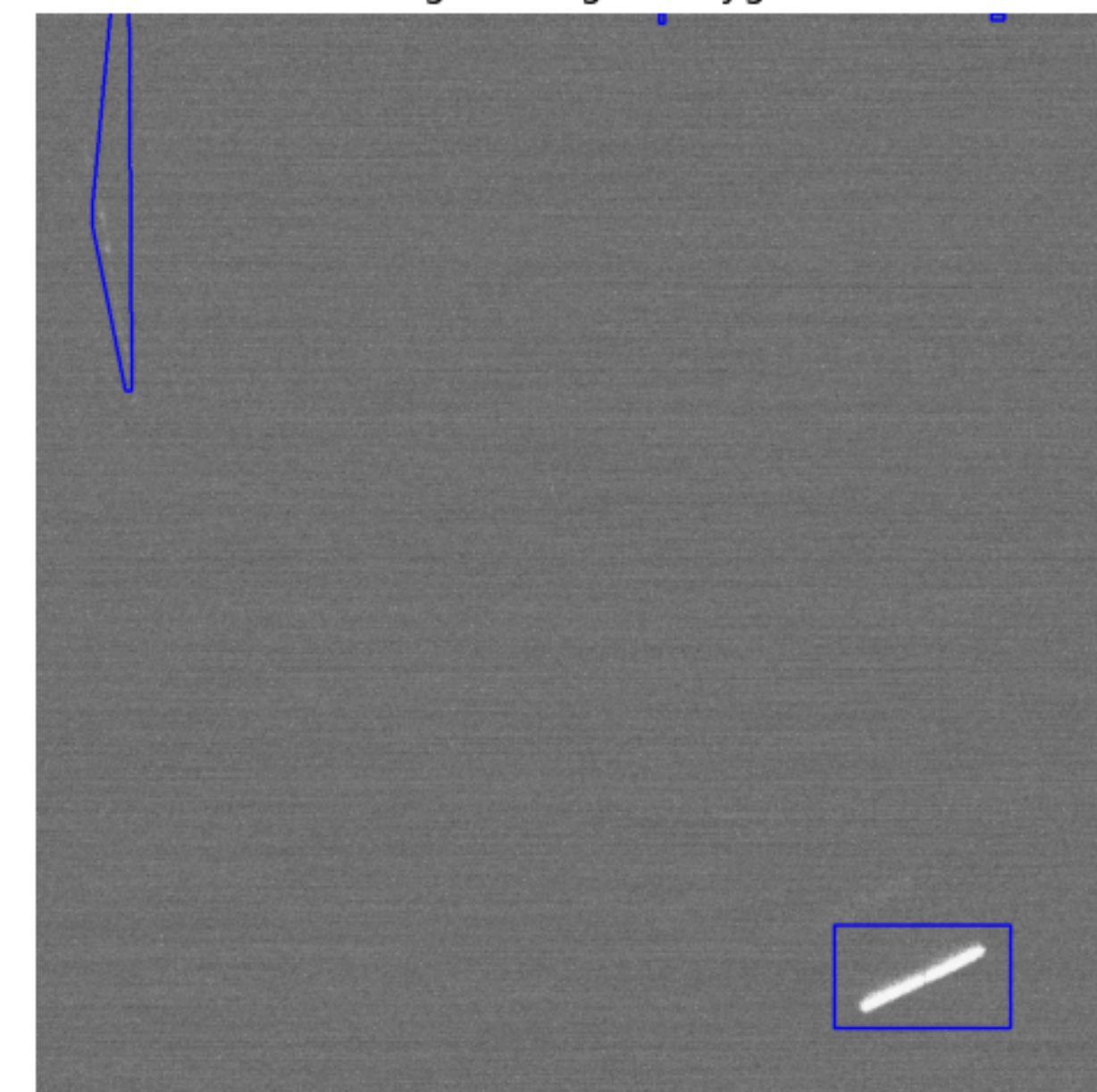
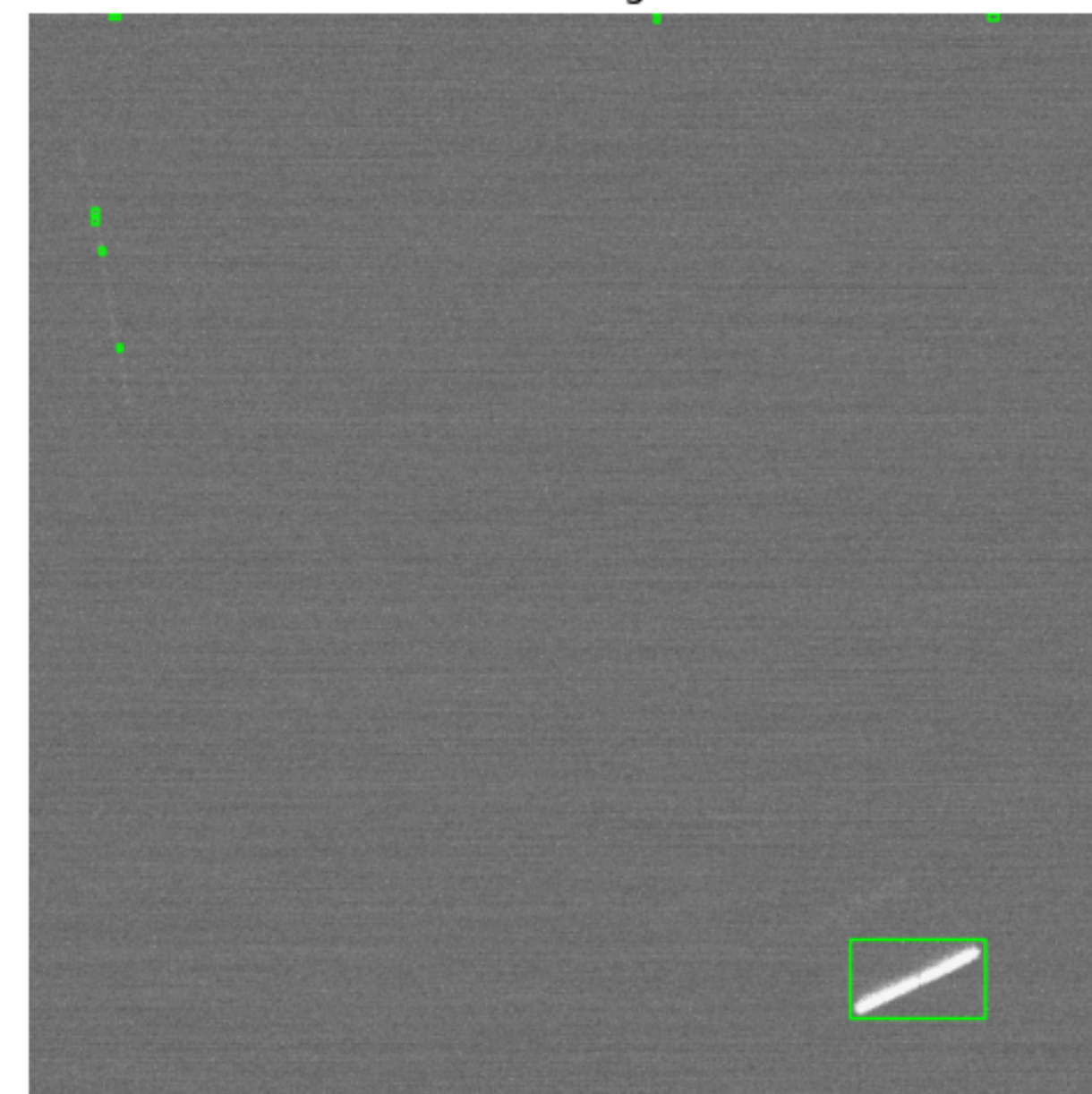
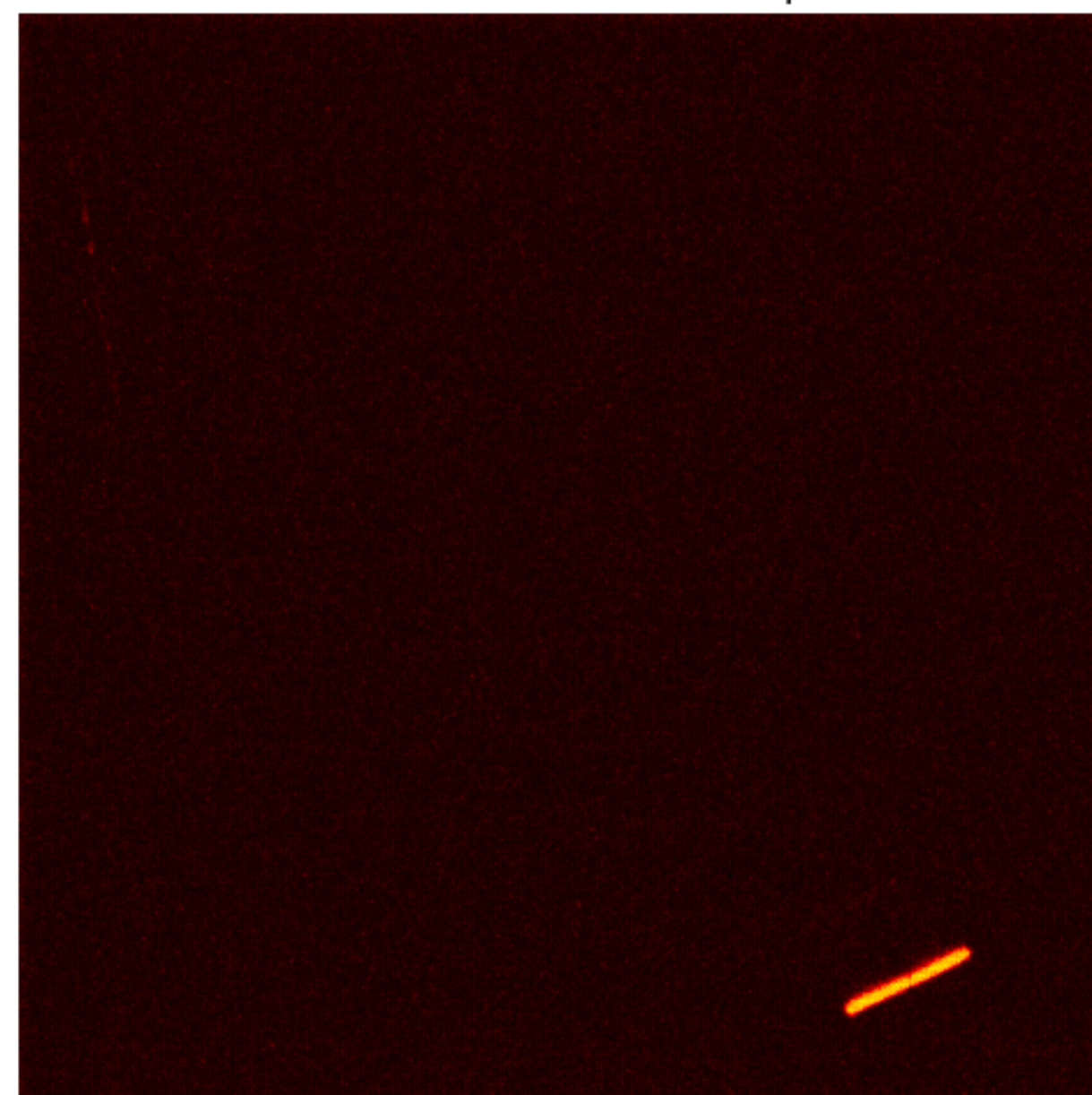
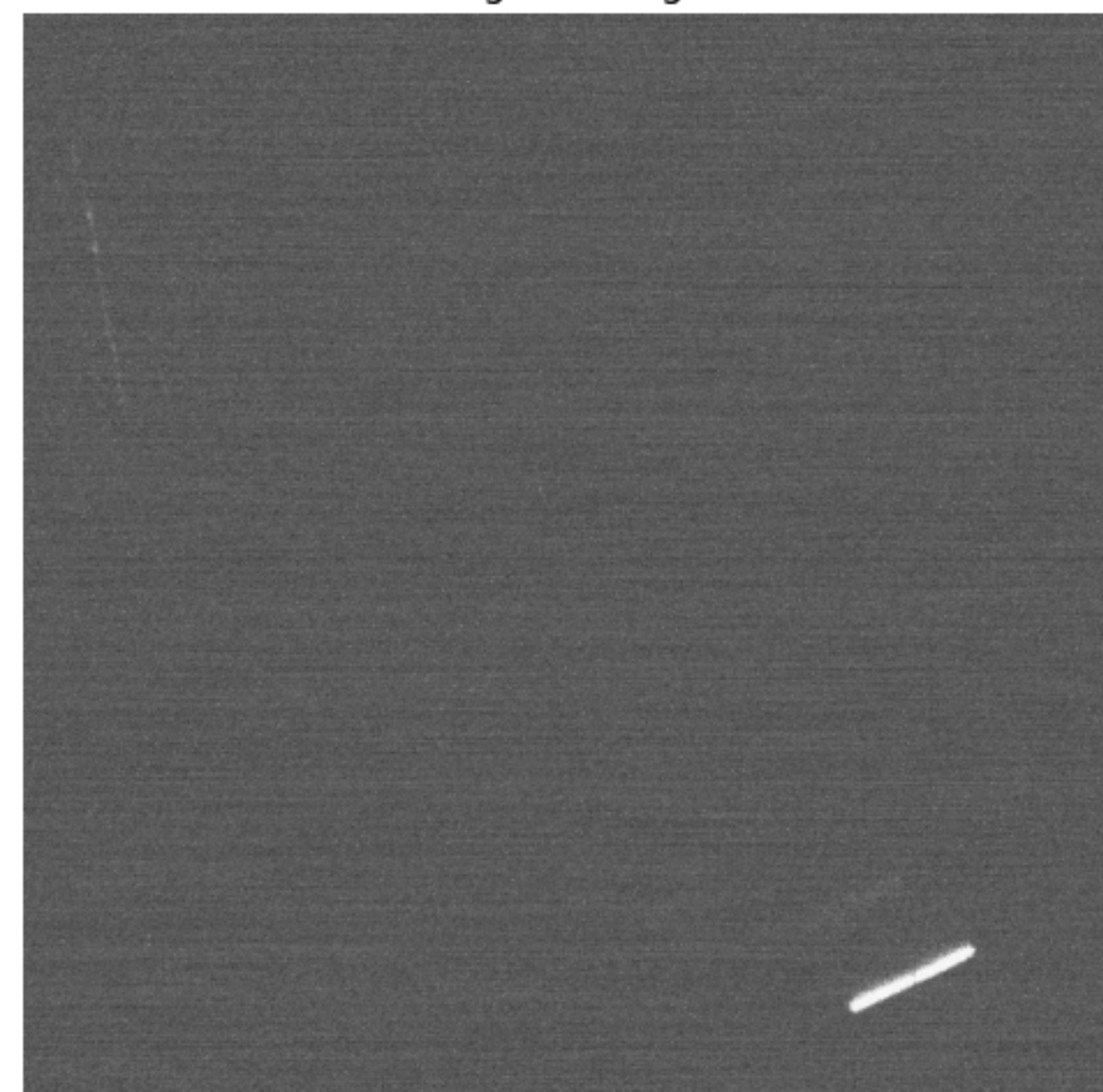


Original Image

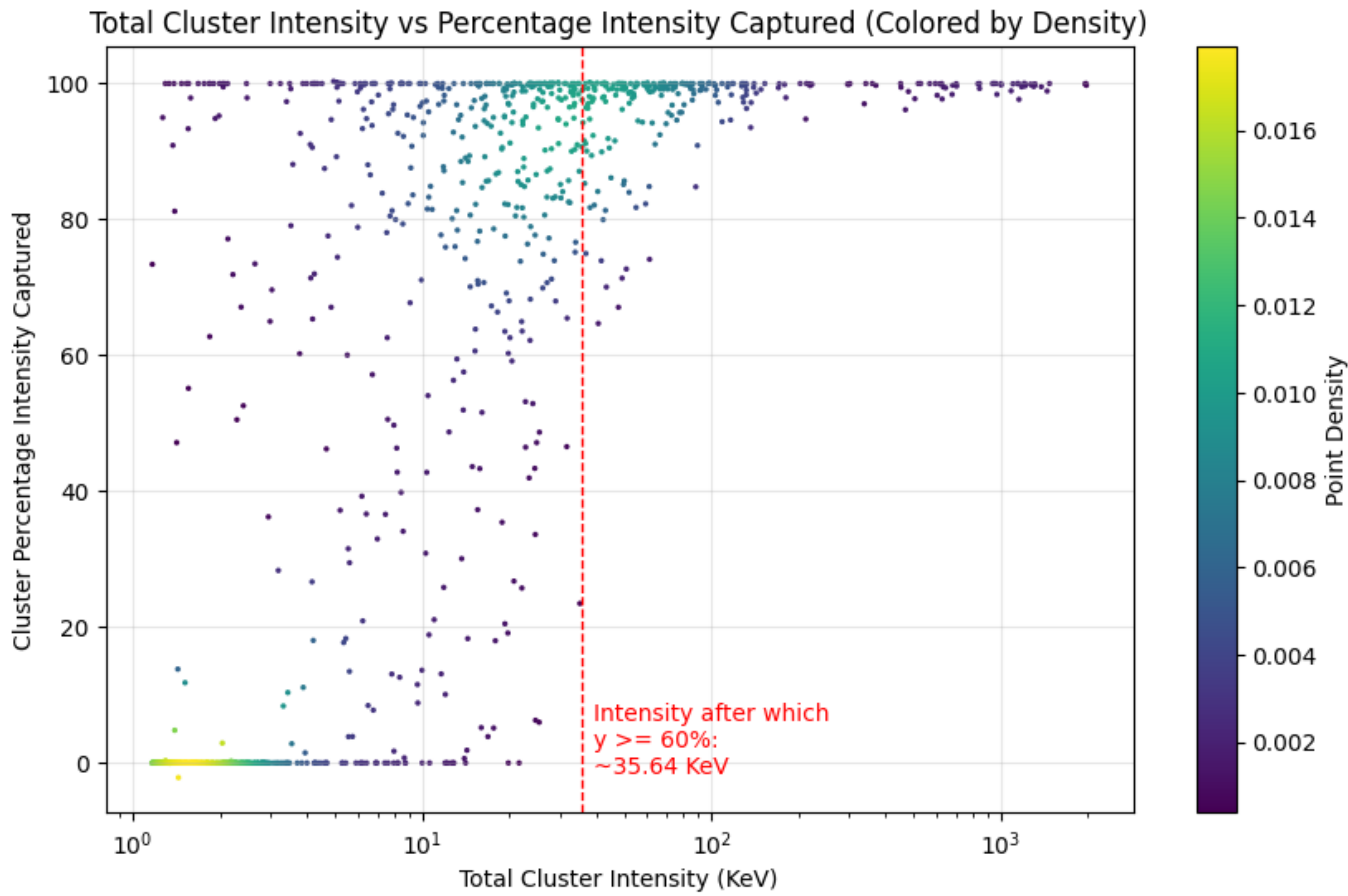
Reconstruction Error Map

Raw Bounding Boxes

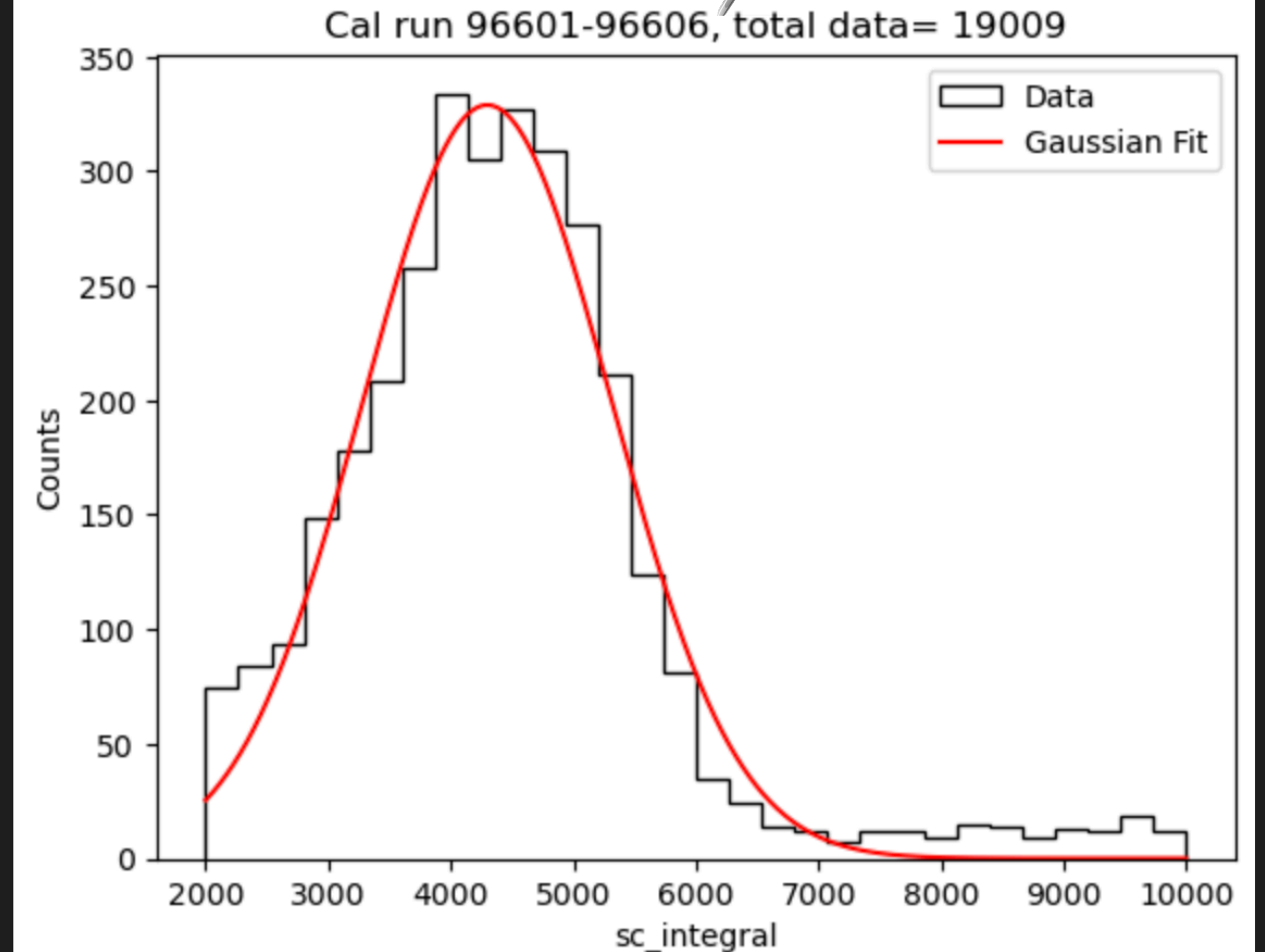
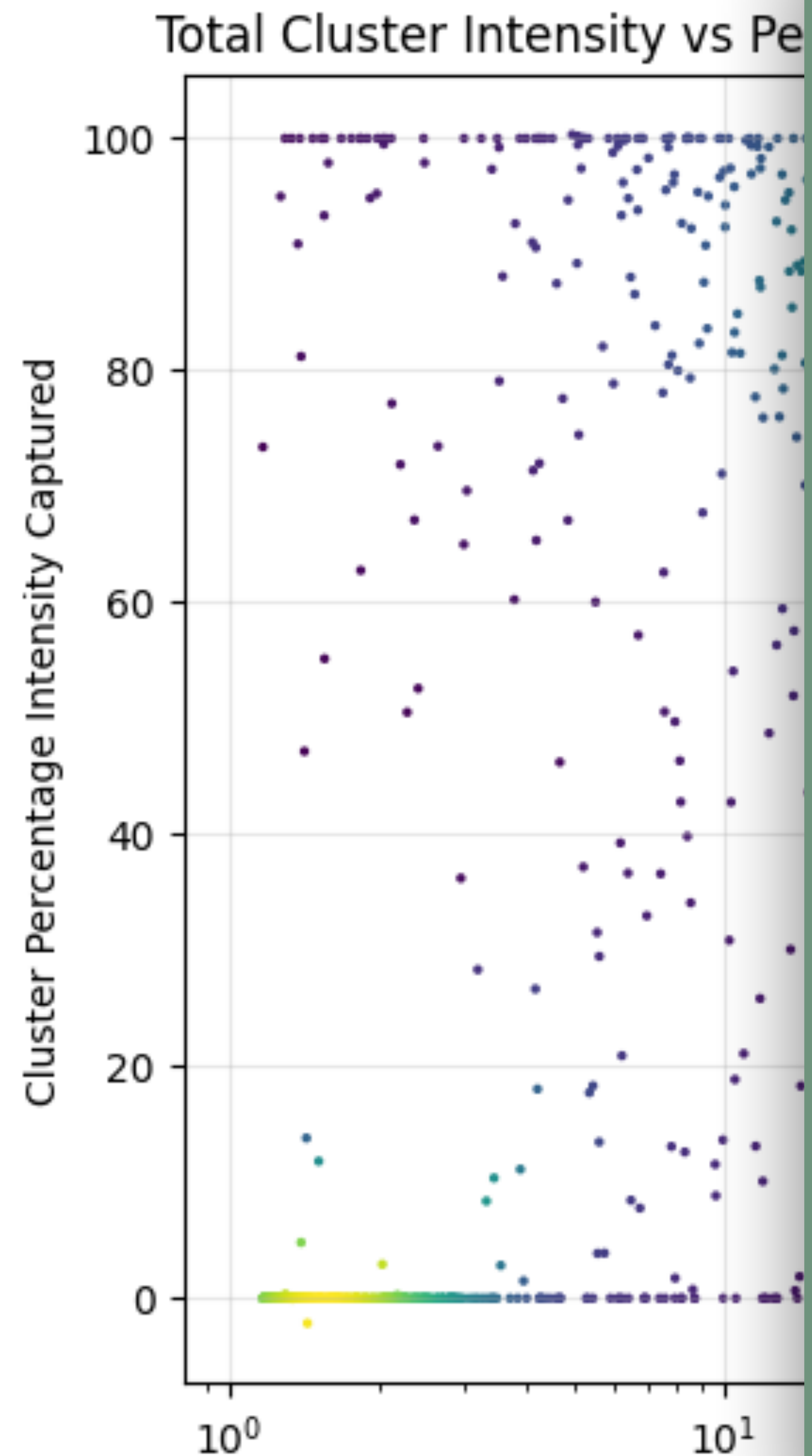
Enlarged Merged Polygons



Redpix Comparison



Redpix Comparison



Fit parameters: a=328.86, x0=4297.12, sigma=1015.31

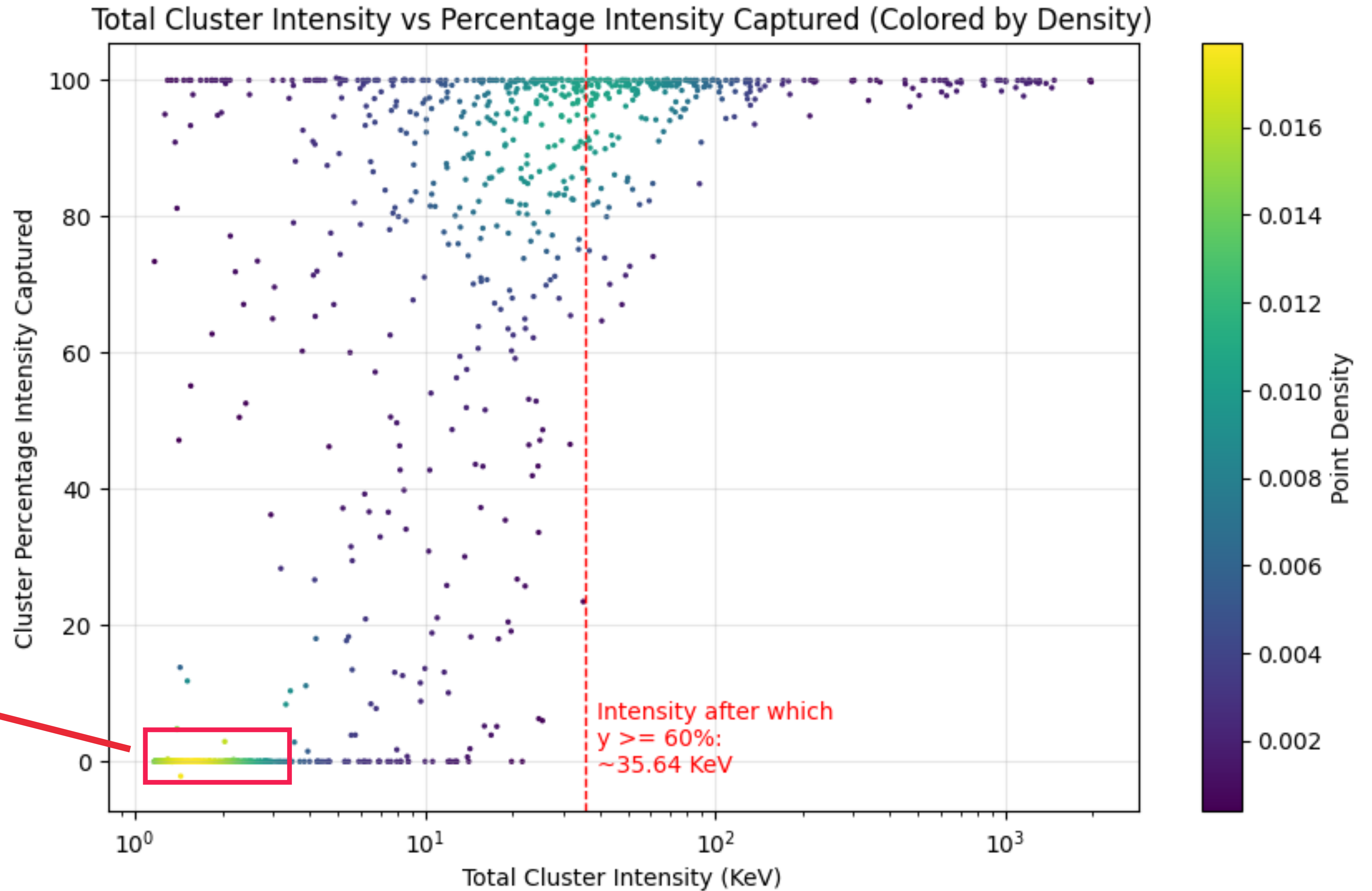
```
conv_to_keV = 5.9/ meanfitt  
print("Conversion factor to keV: ", np.round(conv_to_keV,6))
```

✓ 0.0s

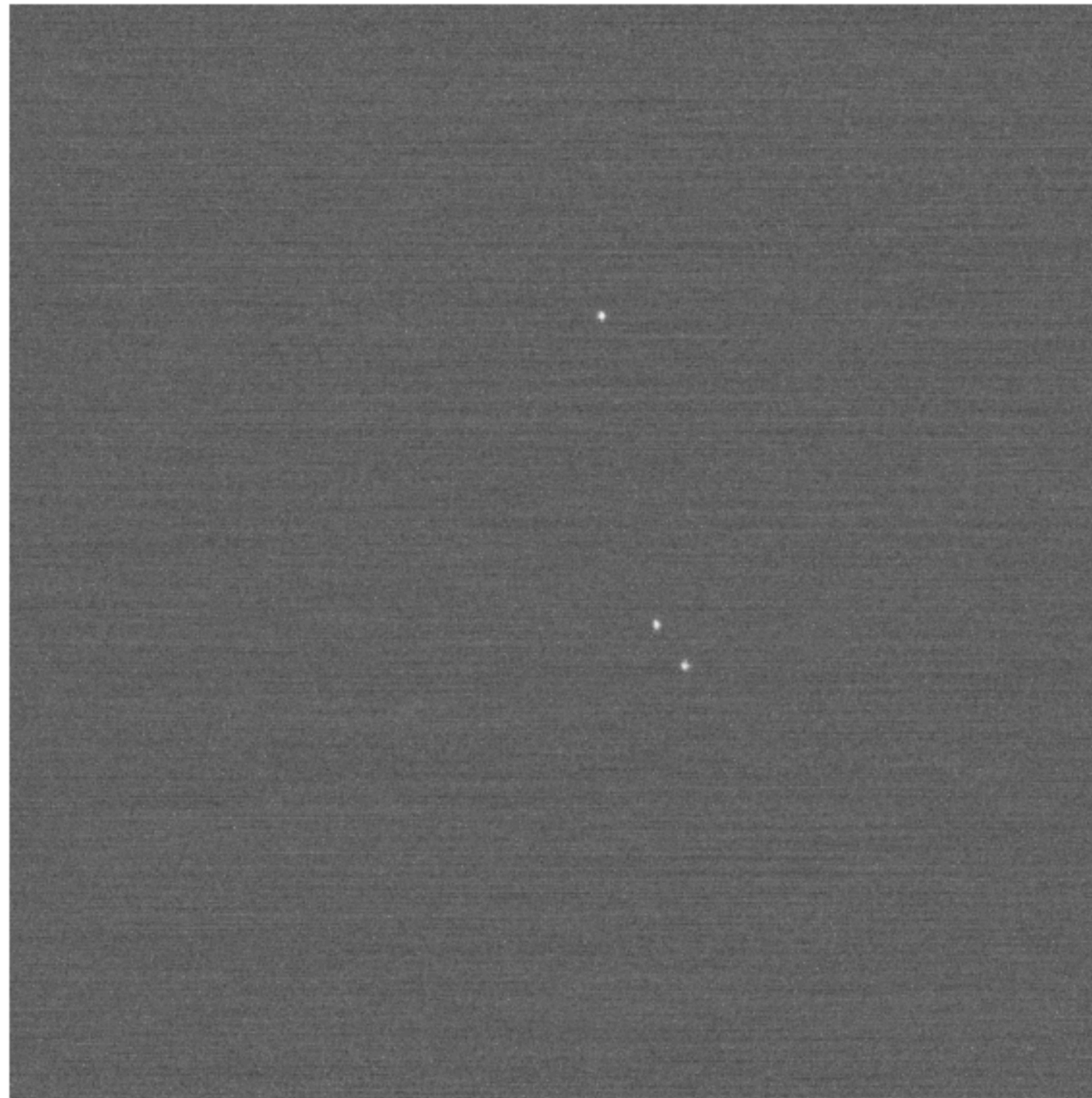
Conversion factor to keV: 0.001373

Redpix Comparison

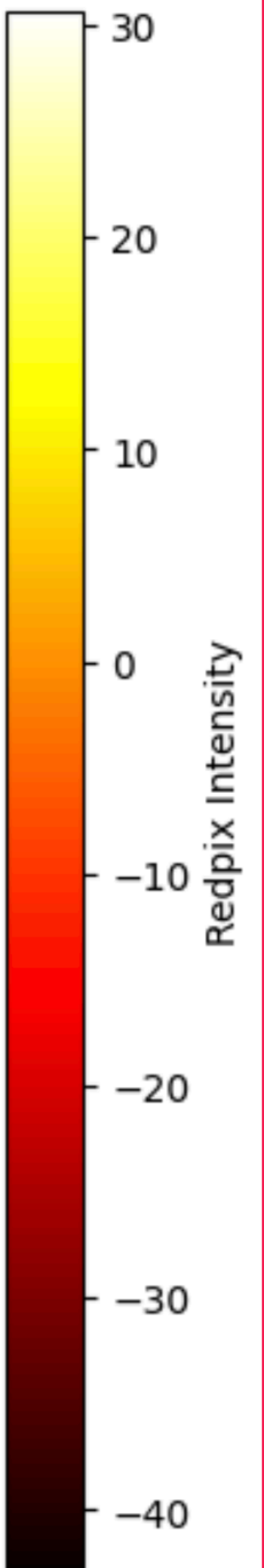
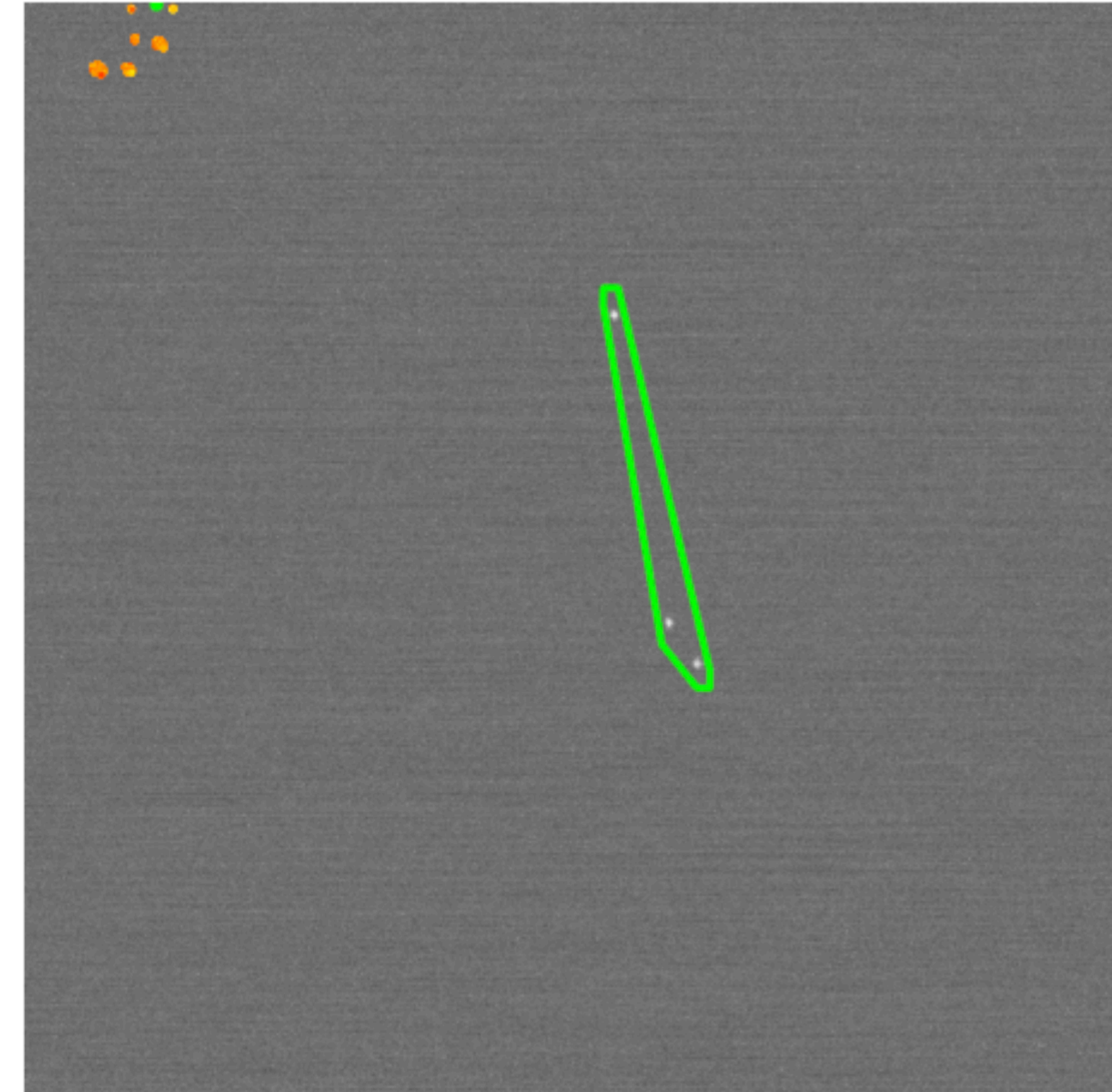
A lot of fake
redpix clusters



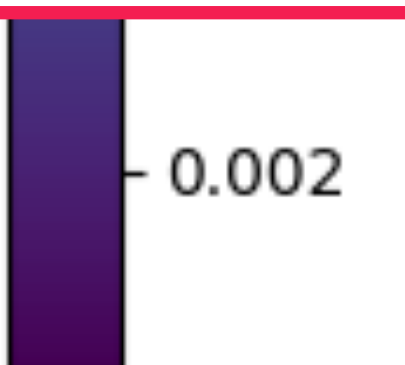
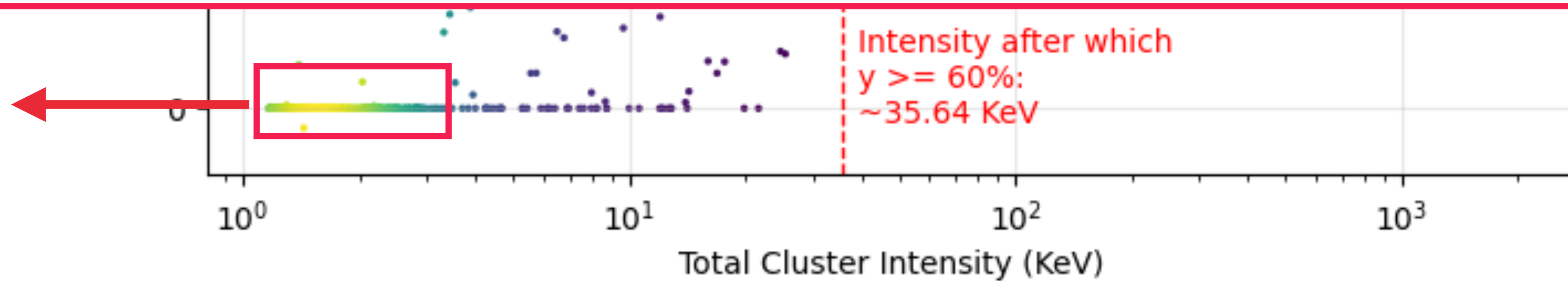
Original: run_96369_event_154



Overlay (Match 58)
Redpix inside: 0.0% | Intensity inside: 0.0%
Area outside polygon: 99.0%



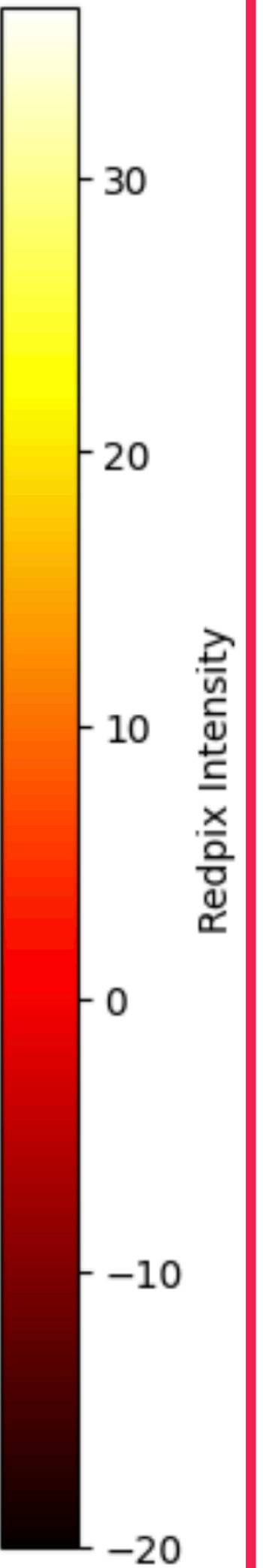
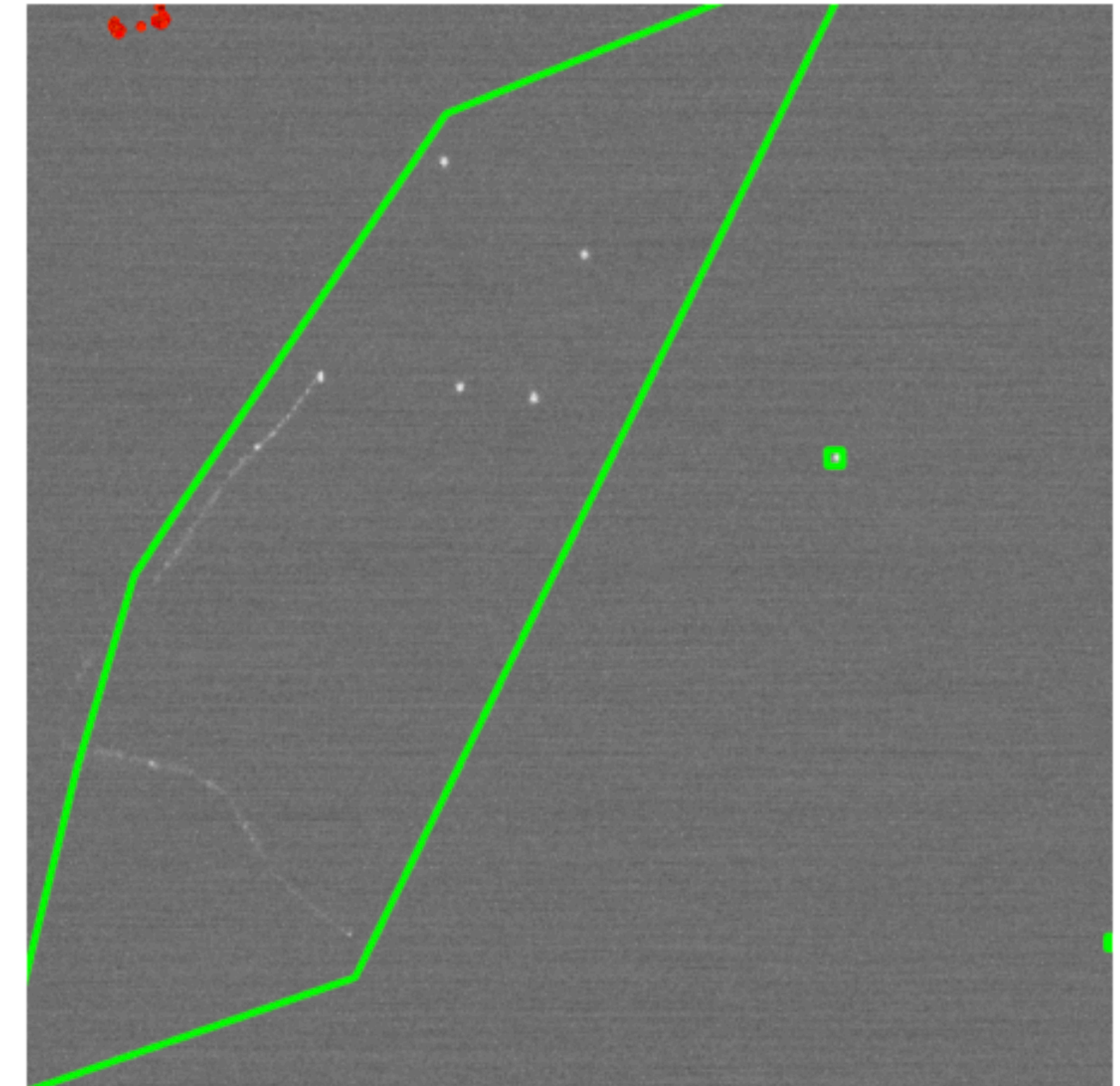
A lot of fake
redpix clusters



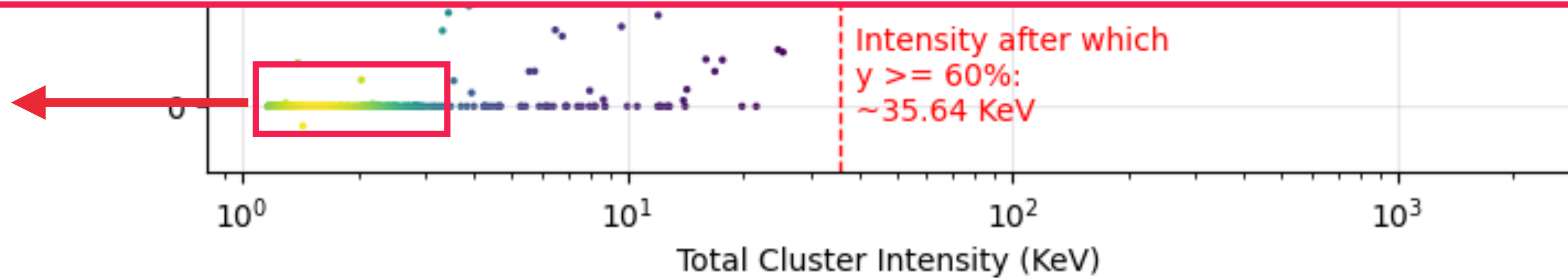
Original: run_96369_event_120



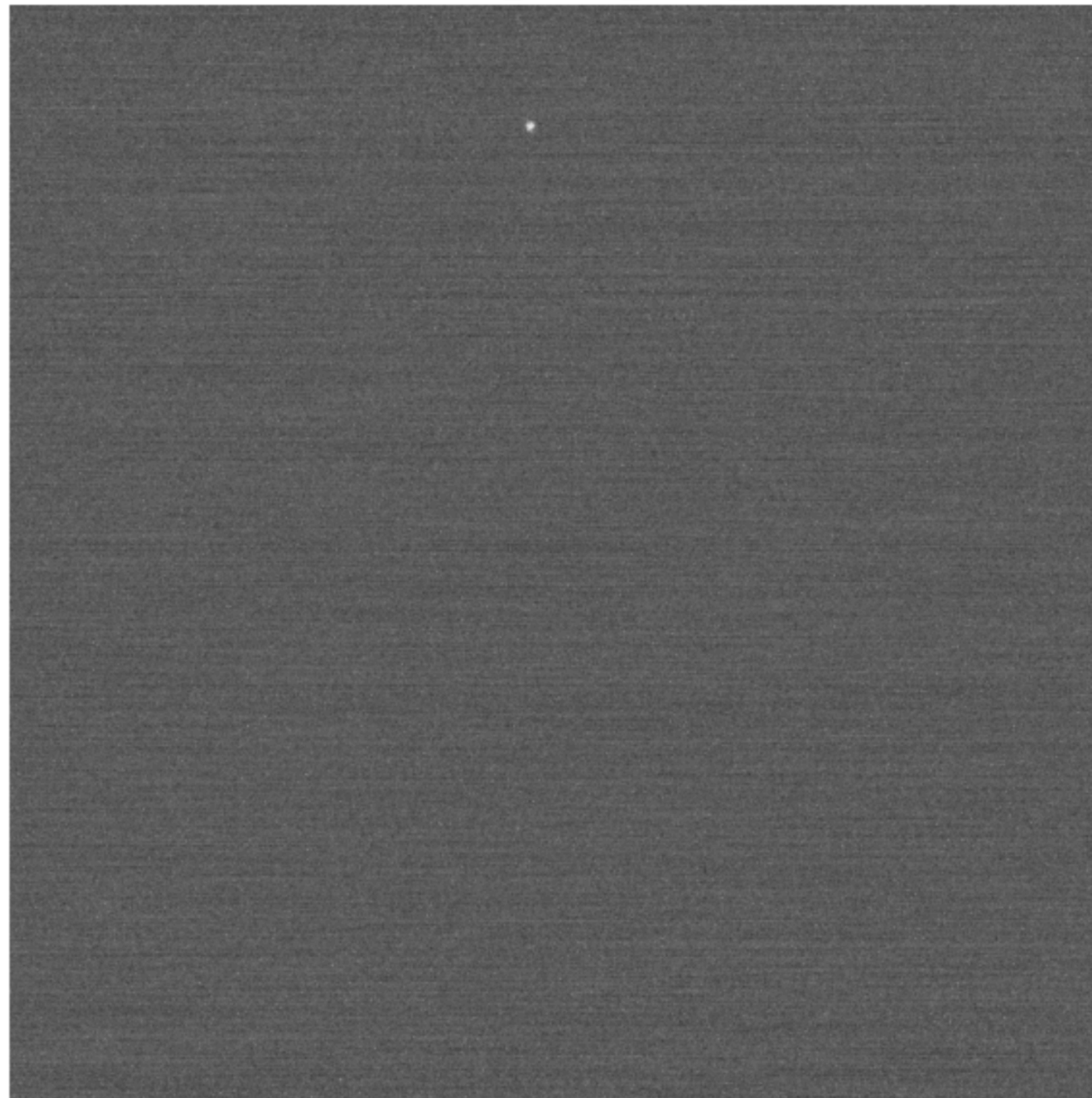
Overlay (Match 57)
Redpix inside: 0.0% | Intensity inside: 0.0%
Area outside polygon: 68.2%



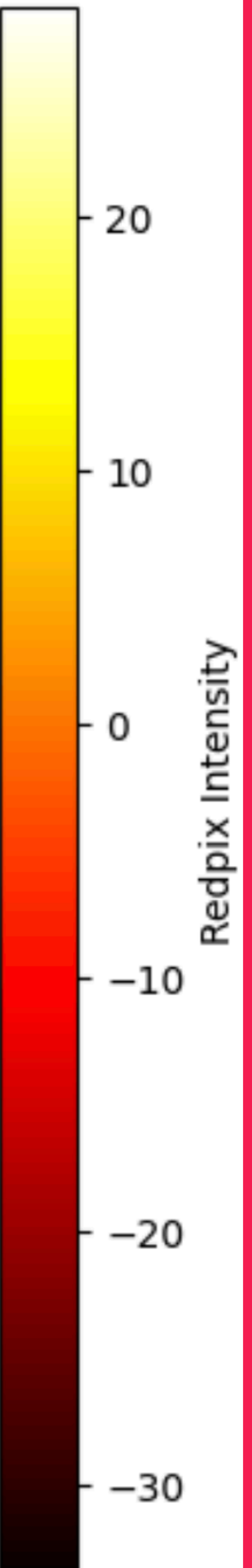
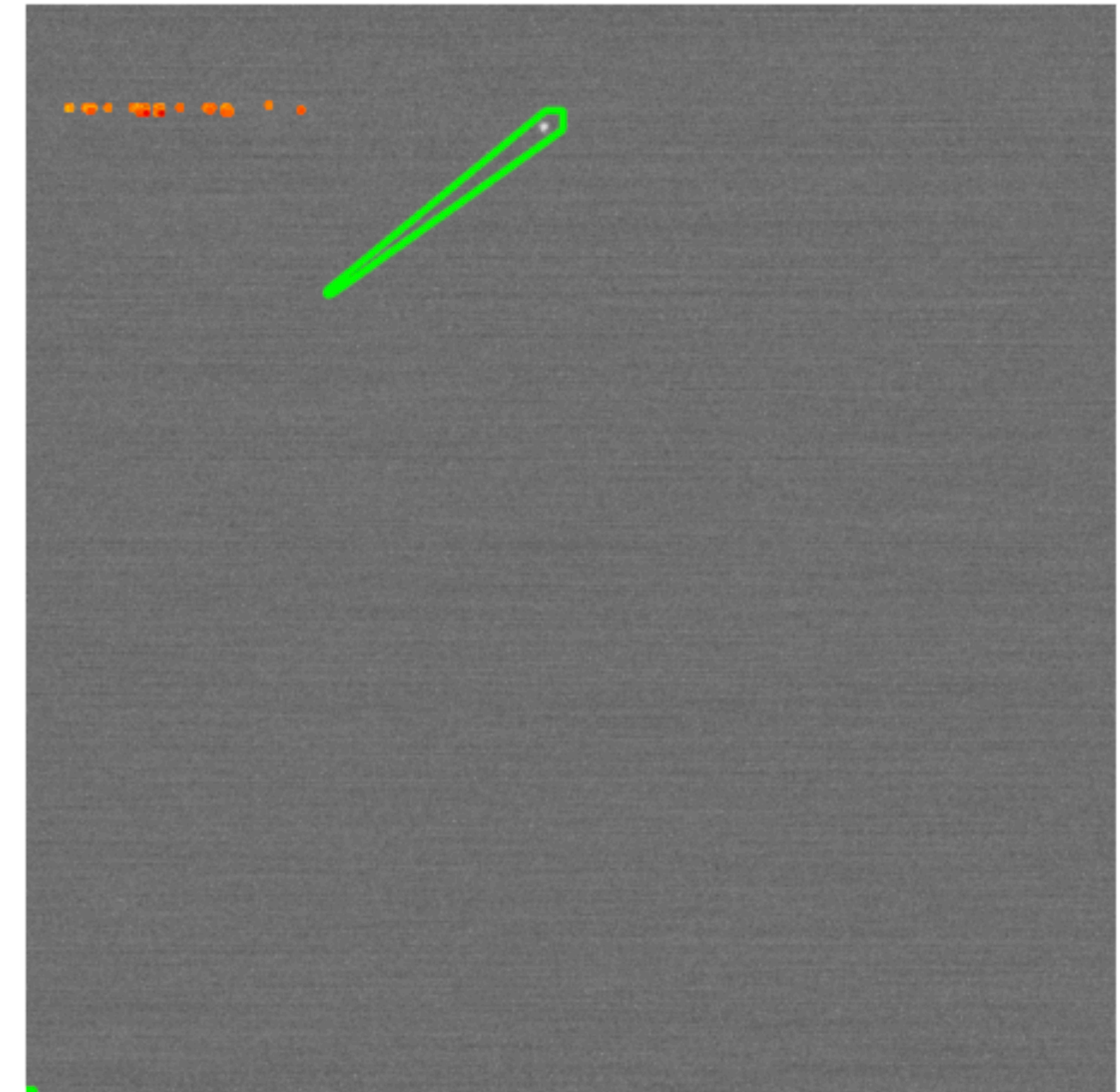
A lot of fake
redpix clusters



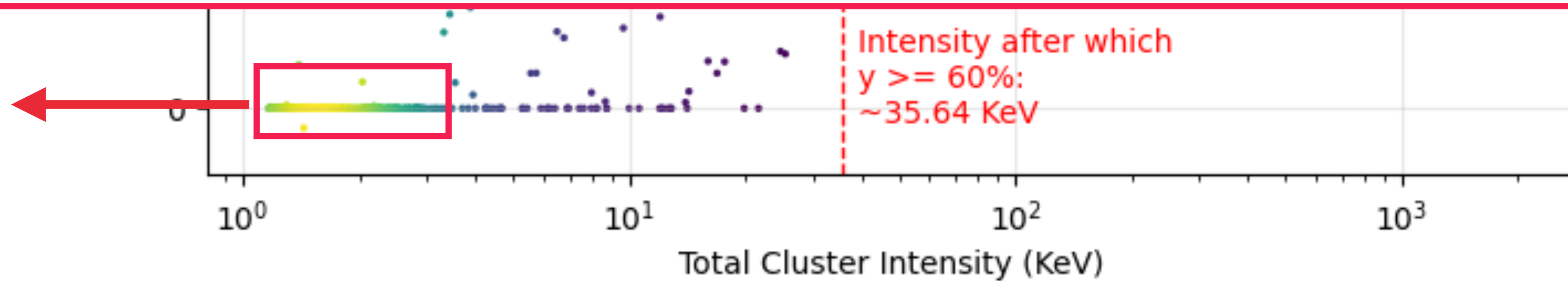
Original: run_96369_event_219



Overlay (Match 60)
Redpix inside: 0.0% | Intensity inside: 0.0%
Area outside polygon: 99.6%



A lot of fake
redpix clusters



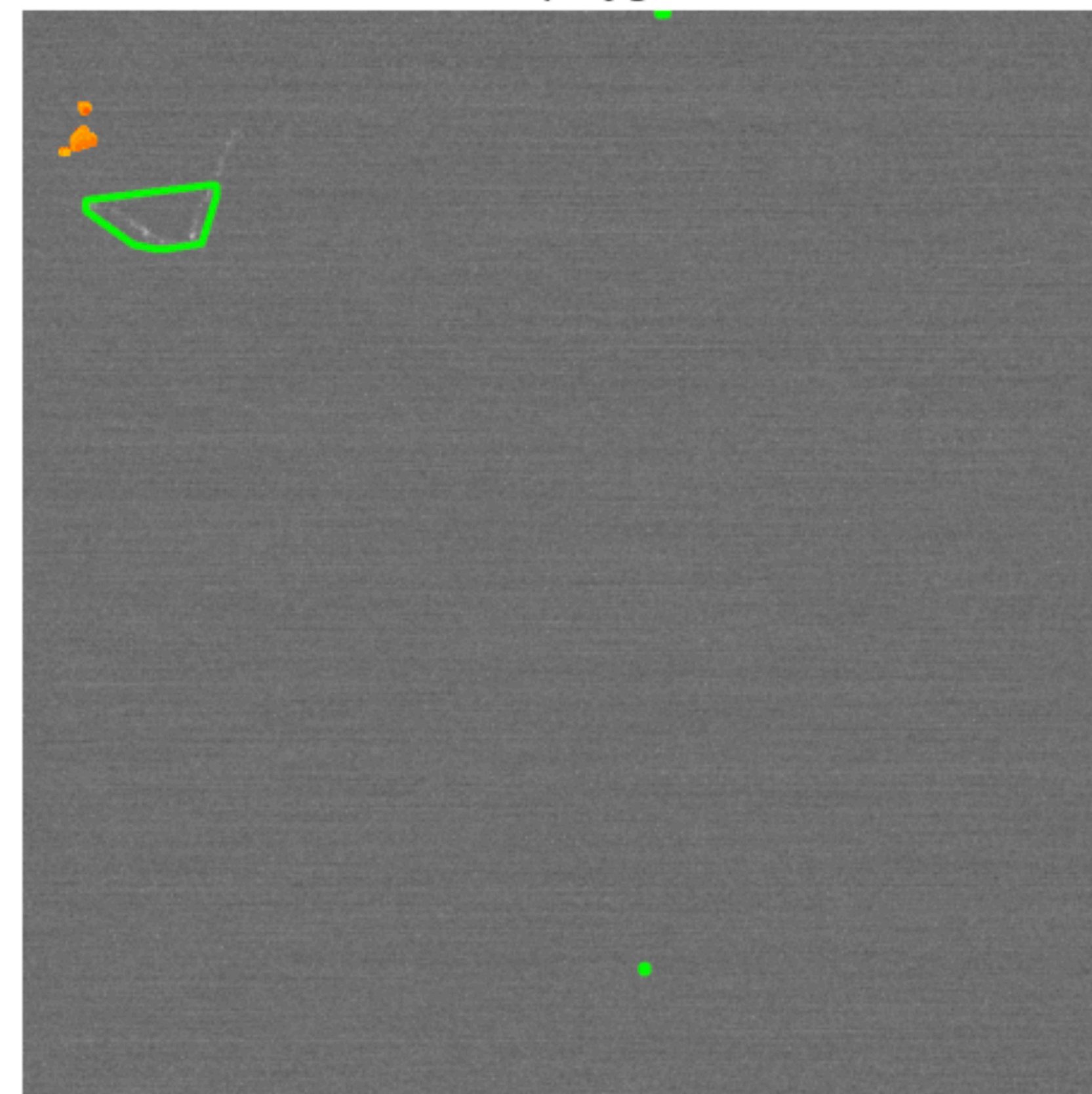
Intensity after which
 $y \geq 60\%$:
~35.64 KeV



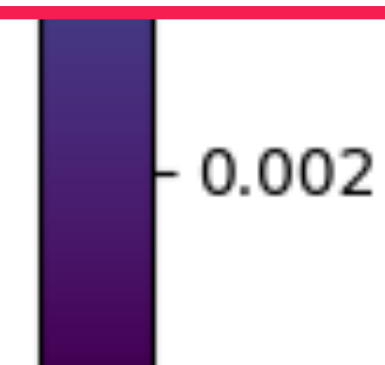
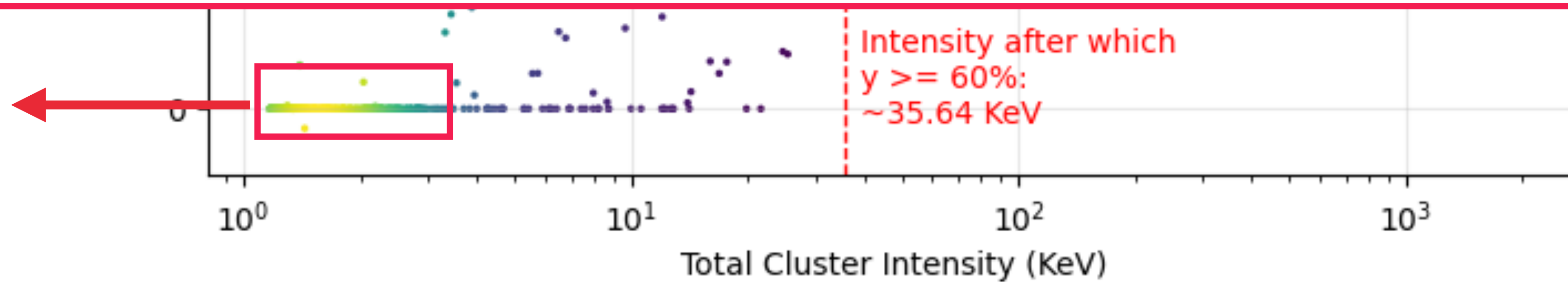
Original: run_96367_event_81



Overlay (Match 11)
Redpix inside: 0.0% | Intensity inside: 0.0%
Area outside polygon: 99.5%



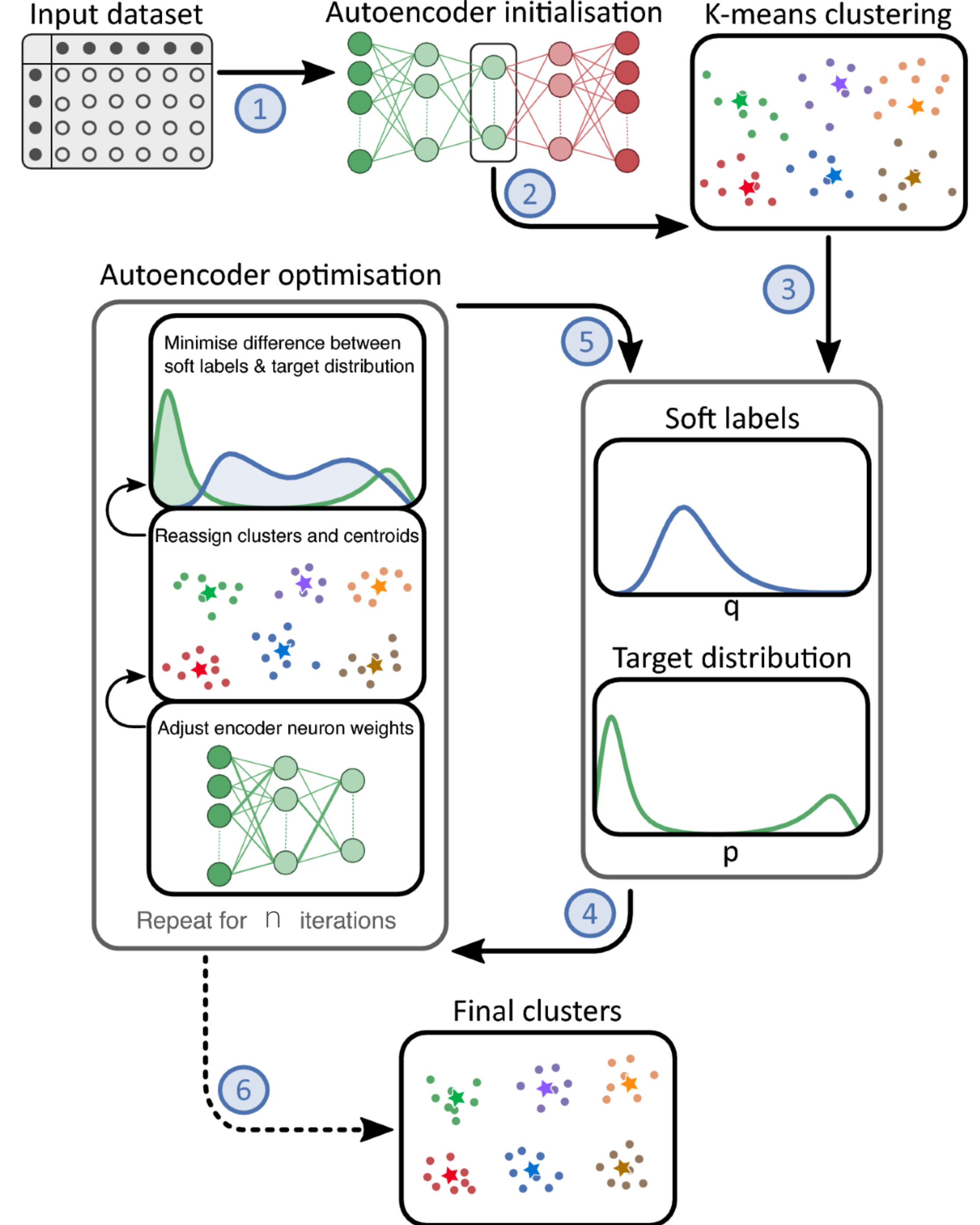
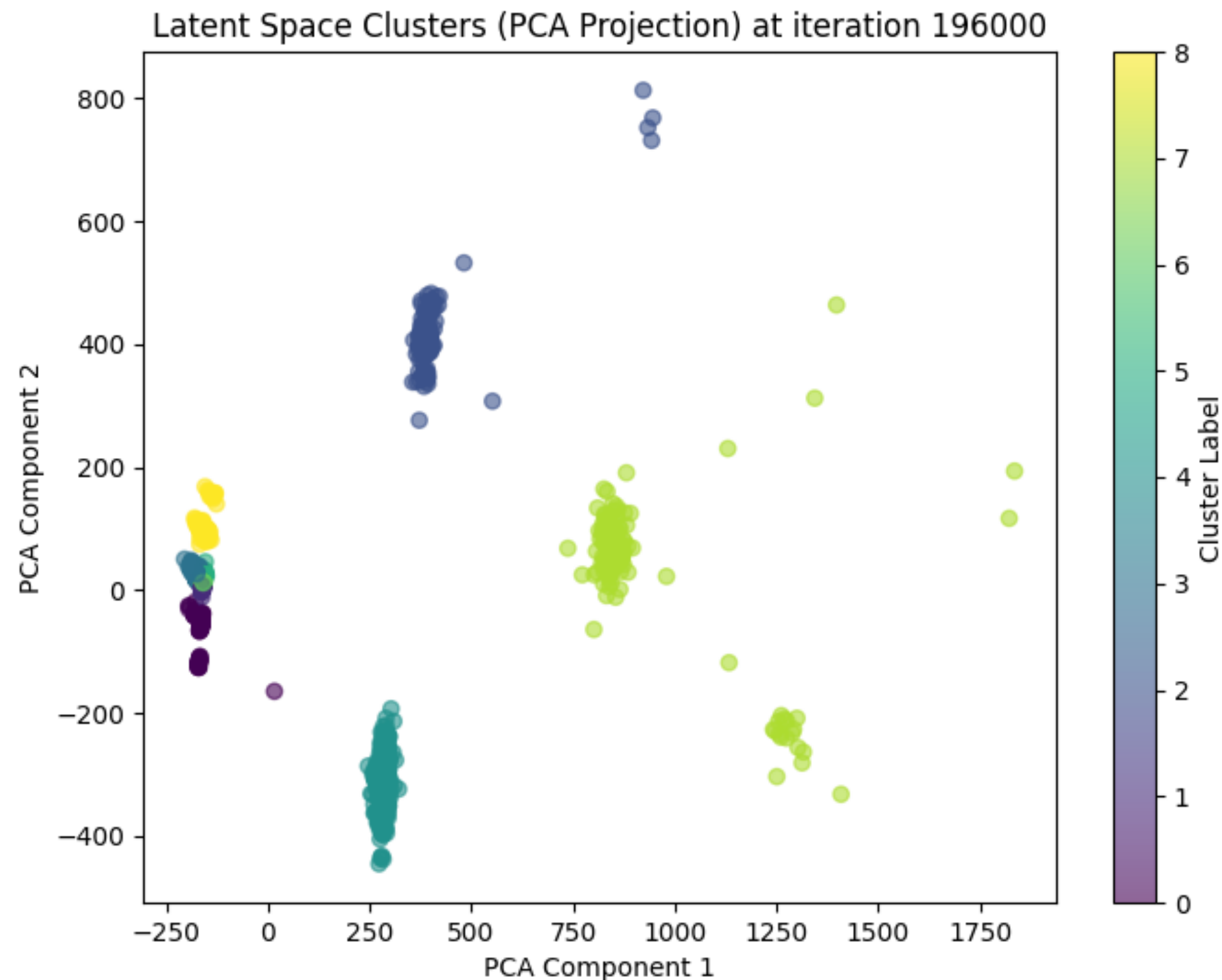
A lot of fake
redpix clusters



Unsupervised Classification

Idea: Deep Embedded Clustering (DEC)

- Assign labels to the clusters based on the latent space representation
- Input: 100x100 patches of the image

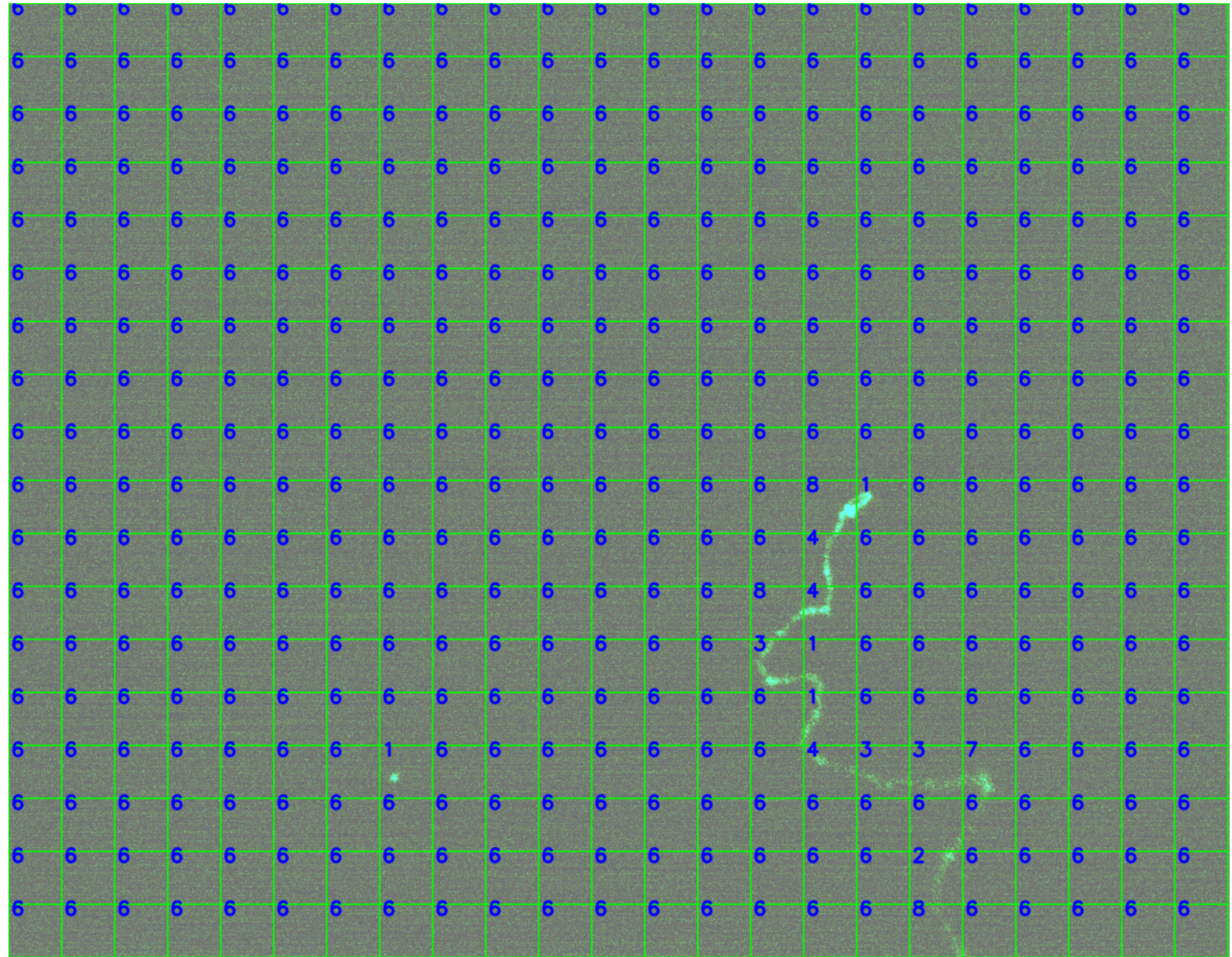


Results

In development

- At the moment the model was trained with a small subset of images
- In order to achieve reasonable results:
- Bigger and more balanced dataset

Original Image with Cluster Labels Overlaid



What's next

- The model is able to identify relevant portions of the images, but redpixes are not a good benchmark
- Defining a benchmark \longrightarrow montecarlo?
- Develop the dataset for the DEC

Thank you for your attention