





# Trigger algorithm based on image processing

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# 1. Introduction

## Motivation

- One of the biggest challenges for the **CYGNO experiment** in the long term will be to manage and store all the data produced by the detector.
  - Each run containing **400 images** need **~1.36 Gb** to be stored.
  - A **single day** of acquisition may produce **~266 Gb** of data.
- The main objective of this work was to study algorithms capable of **distinguishing** which images contain any kind of **signal** or only **electronic noise.**
- This proposal was called **image-based trigger algorithm.**

## What was done

- Two algorithms were proposed:
  - Based on **filtering.**
  - Based on **CNN.**

#### • Comparative analysis:

- **Detection performance** on low energy simulated signals.
- **Time** analysis.
- Comparison with the **reconstruction algorithm.**

# 2. Algorithms









# **CNN** architecture



# 3. Results

## Datasets

## • Training:

- **Noise dataset:** 600 images from pedestal runs (Run 4 underground).
- **ER and NR signal simulation:** 600 images each containing 0.25-1 keV signals added to pedestal runs (different from noise dataset).

### • Validation:

- Noise dataset: 200 images from pedestal runs.
- **ER and NR signal simulation:** 200 images each containing 0.25-1 keV signals.

#### • Test:

• Same configuration as validation.

# Trigger performance

- The CNN can detect 80% of the 0.25 keV NR and ER from the test dataset with a ~0.5% false alarm.
- The Gaussian filter would have ~10% false alarm to have the same detection performance.
- Both methods outperforms the reconstruction code in detecting 0.25 keV signals.
- All methods can **easily detect** signals with energies above **0.5 keV.**



## **Processing time**

- The Gaussian filter needs ~0.25 and ~0.02 s per event using CPU<sup>1</sup> and GPU<sup>2</sup> respectively.
- The CNN needs ~0.55 and ~0.2 s per event using CPU and GPU respectively.
- A higher detection performance is compensated with a slower processing time.



<sup>1</sup>CPU: Notebook01 cloud <sup>2</sup>GPU: Tesla T4 Google Collab

# 4. Conclusions

## Conclusions

- The results show that the **trigger algorithms** based on the **Gaussian filter** and **CNN** can achieve a **80% signal detection** rate on **0.25 keV ER** and **NR** simulated events with a **10%** and **0.5% false alarm** rate respectively.
- The CNN algorithm needs a GPU to have a proper time margin to predict the data, whereas the Gaussian filter may be used with a CPU.
- All the signals detected by the reconstruction algorithm were detected by the trigger algorithms.

## Next steps

- Study methods to simplify a trained CNN model: Bit reduction, weight combination, pruning and vectorization.
  - First attempt did not work, on going.
- Apply the CNN on the DAQ machine.
  GPU: Quadro RTX 5000
- Test popular CNN architectures such as AlexNET, GoogleLeNet, Unet with necessary adaptations.
- Write a paper based on these results.

# Thank you

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