

# FCAL Test Beam Data and Geant4 Simulations



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On behalf of the FCAL collaboration



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# Outline

- Introduction
- Beam test design in simulations
- LumiCal calibration
- Clustering algorithms and position reconstruction
- Summary

# LumiCal in LC Experiments

## Goals:

- Precise integrated luminosity measurements;
- Extend a calorimetric coverage to small polar angles. Important for physics analysis.

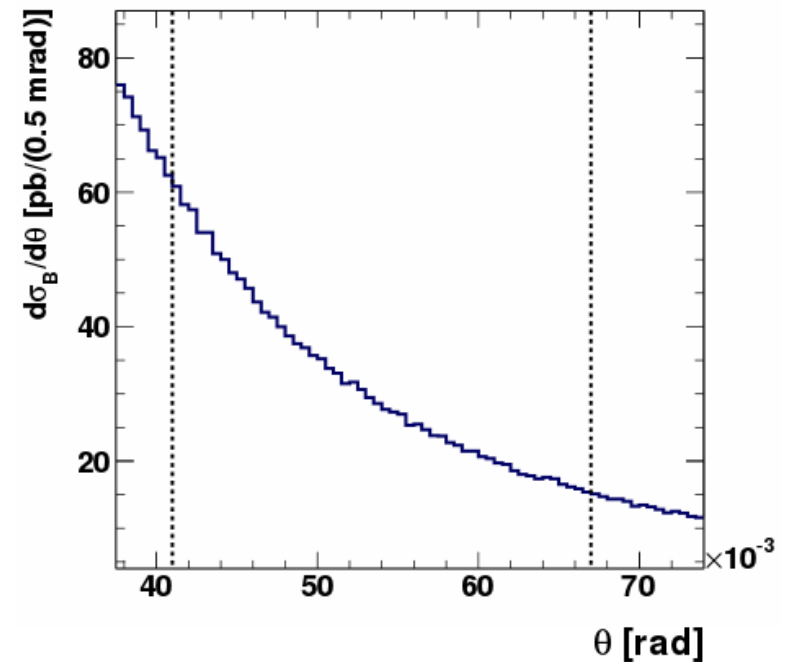
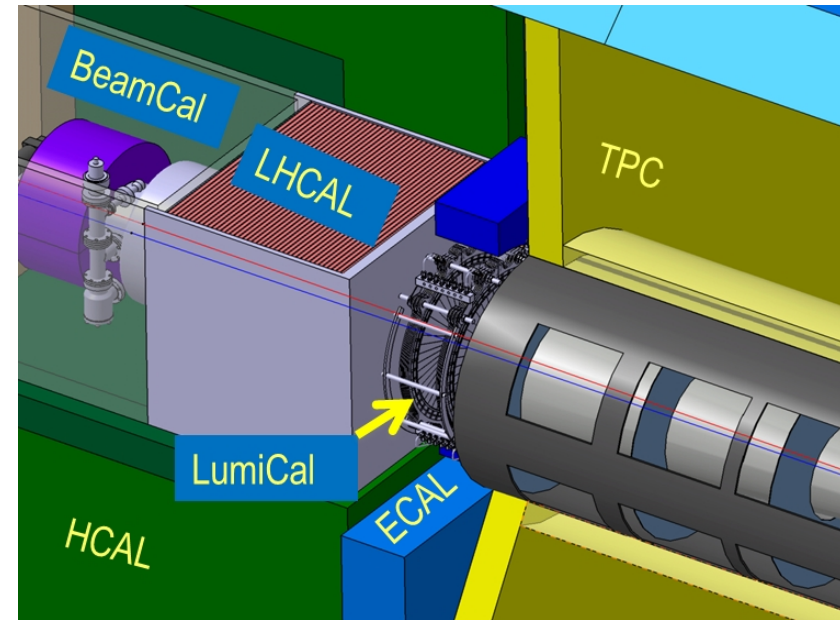
## LumiCal Design:

- Electromagnetic sampling calorimeter;
- 30 layers of 3.5 mm thick tungsten plates with 1 mm gap for silicon sensors;
- symmetrically on both sides at  $\sim 2.5\text{m}$  from the interaction point.

**Luminosity measurement:** 
$$L = \frac{N_B}{\sigma_B}$$

$N_B$  – Bhabha events in a certain polar angle ( $\theta$ );

$\sigma_B$  – integral of the differential cross section over the same  $\theta$  range.



# LumiCal Geometry

Uncertainty in luminosity measurement depends on the polar angle bias  $\Delta\theta$  and minimum polar angle  $\theta_{\min}$  as:

$$\left(\frac{\Delta L}{L}\right)_{\text{rec}} \approx 2 \frac{\Delta\theta}{\theta_{\min}} \quad \Delta\theta \text{ depends on polar angular pad size } l_{\theta}.$$

For  $l_{\theta} = 0.8 \text{ mrad}$ ,  $\Delta L/L = 1.6 \cdot 10^{-4}$ .

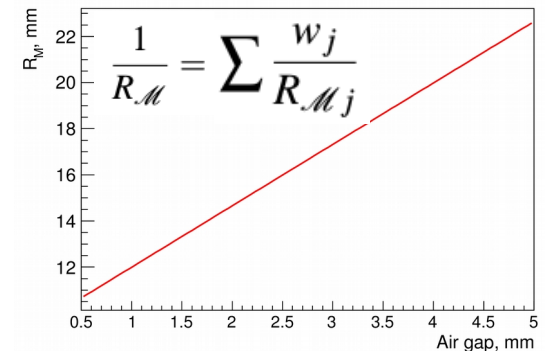
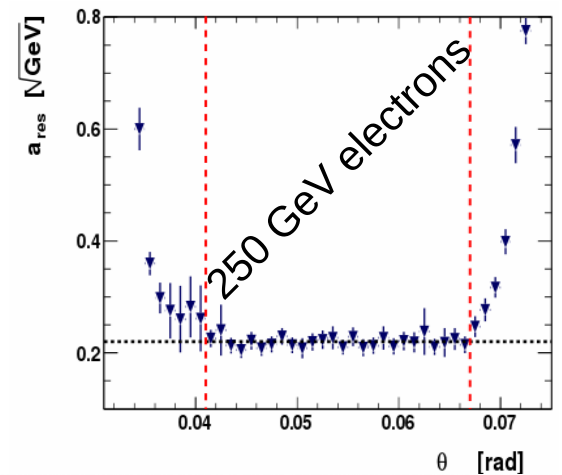
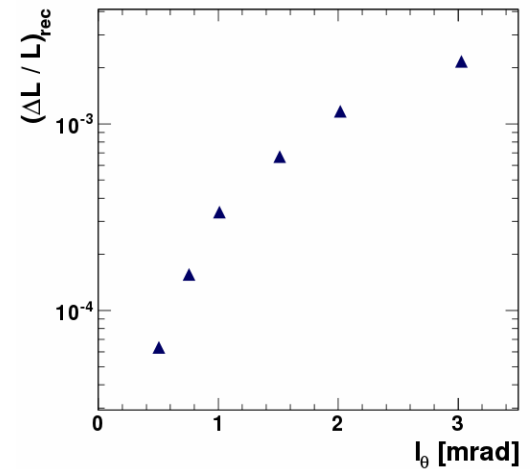
Energy resolution: 
$$\frac{\sigma_E}{E} = \frac{a_{\text{res}}}{\sqrt{E_{\text{beam}} \text{ (GeV)}}},$$

$$a_{\text{res}} = (0.21 \pm 0.02) \sqrt{\text{GeV}}.$$

LumiCal fiducial volume:  $41 < \theta < 67 \text{ mrad}$

$R_M$  as function of the air gap between 3.5 mm thick tungsten plates

Reducing air gap from 4.5 mm to 1 mm gives  $R_M$ : 21 mm  $\rightarrow$  12 mm.

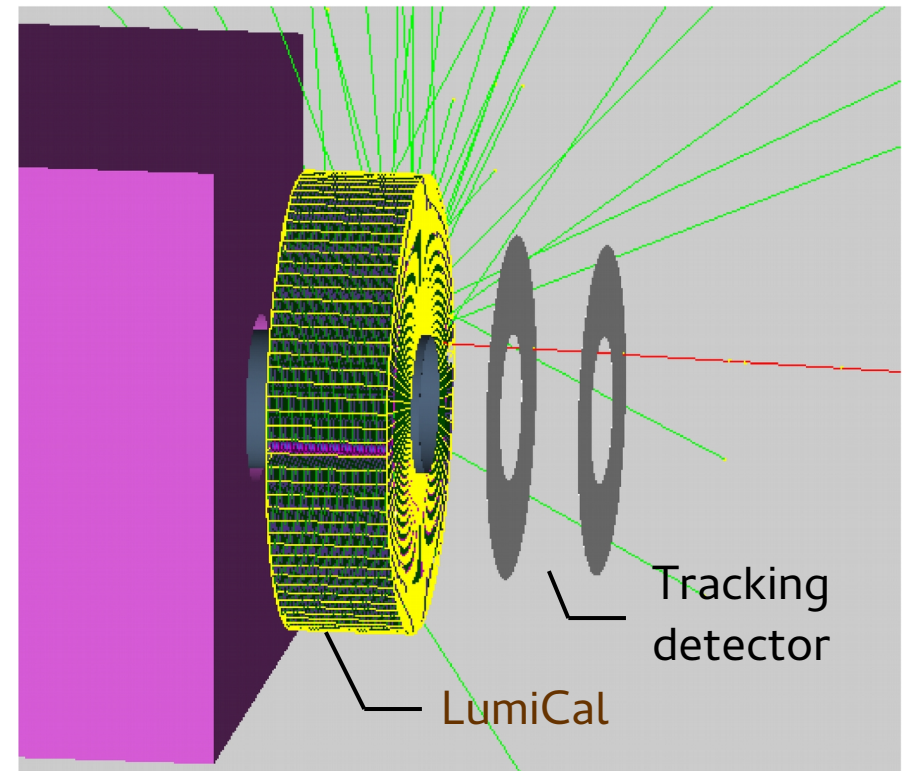


# Tracking Detector in Front of LumiCal

- Improve polar angle measurement accuracy;
- LumiCal alignment;
- Provide more information to enable  $e/\gamma$  identification, important for various physics studies, e. g. photon structure function study, important for BSM searches.

## Studied in Simulations

- Modified versions of LuCaS (Geant4 app. For LumiCal simulation);
- Two layers of Si sensors with different thickness and distance to LumiCal;
- No negative affects on reconstruction in LumiCal;
- High efficiency of  $e/\gamma$  identification for tracker with submillimeter position resolution for single  $e$  or  $\gamma$  event.



Study  $e/\gamma$  identification in beam test

# Beam Test Goals

## DESY test beam facilities:

- Electron beam 1 – 6 GeV;
- Dipole magnet 1 – 13 kGs;
- EU Telescope with 6 planes of Mimosas26 detectors;

## Performance of the compact LumiCal prototype:

- Detector modules performance: noise, saturation, S/N, etc;
- Energy response to  $e^-$  beam of 1 – 6 GeV;
- Electromagnetic shower development study, Moliere Radius measurement.

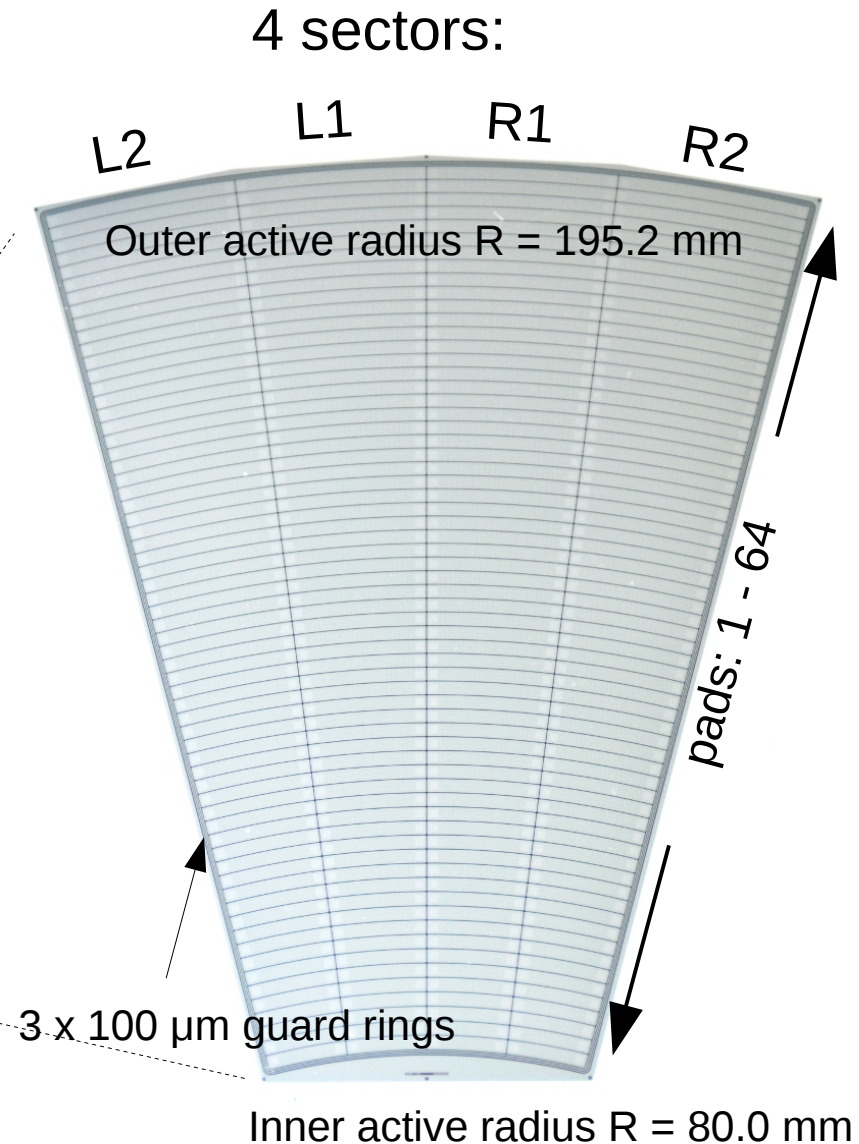
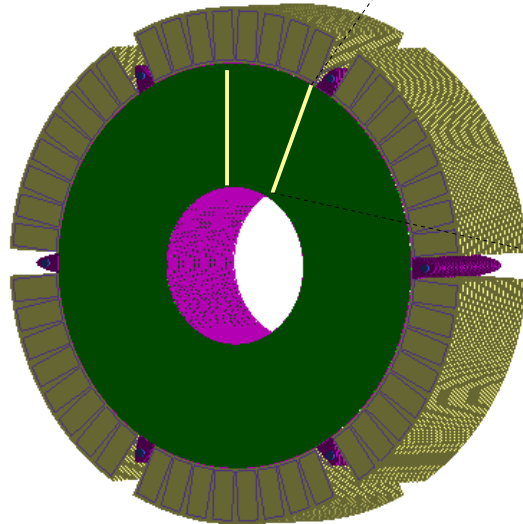
## $e/\gamma$ identification with tracking detector in front of LumiCal:

- Back scattering as a function of distance from LumiCal;
- Identification efficiency.

# LumiCal and Sensor Design

- Silicon sensor
- thickness 320  $\mu\text{m}$
- DC coupled with readout electronics
- p+ implants in n-type bulk
- 64 radial pads, pitch 1.8 mm
- 3 guard rings
- 4 azimuthal sectors in one tile, each 7.5°
- 12 tiles make full azimuthal coverage

LumiCal thickness:  
30X<sub>0</sub> → 13.5 cm.





# Saturation in Readout with SRS and APV-25

Next generation of LumiCal electronics is under development and will be available in 2017.

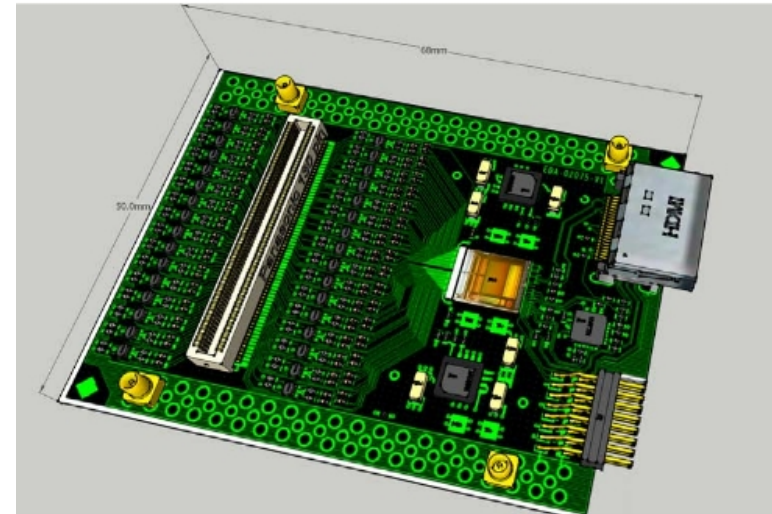
## Temporary alternative solution:

Front-end chip APV25:

- Designed for CMS silicon microstrip detectors (used for Belle II SVT);
- 128 channels;
- Shaping time (min): 50 ns;
- Supports both signal polarities;
- Sampling rate 40 MHz;
- Supported by SRS;
- Available at CERN stock.

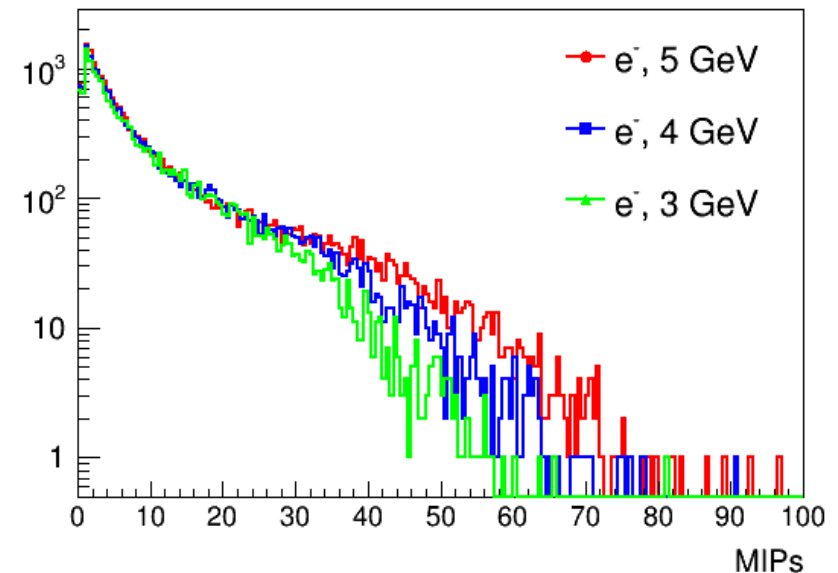
The APV-25 range in case of LumiCals sensor:  $\sim 8$  MIPs

Additional circuit: "charge divider" - could help to avoid saturation.



Front-end board (hybrid) with APV25 chip

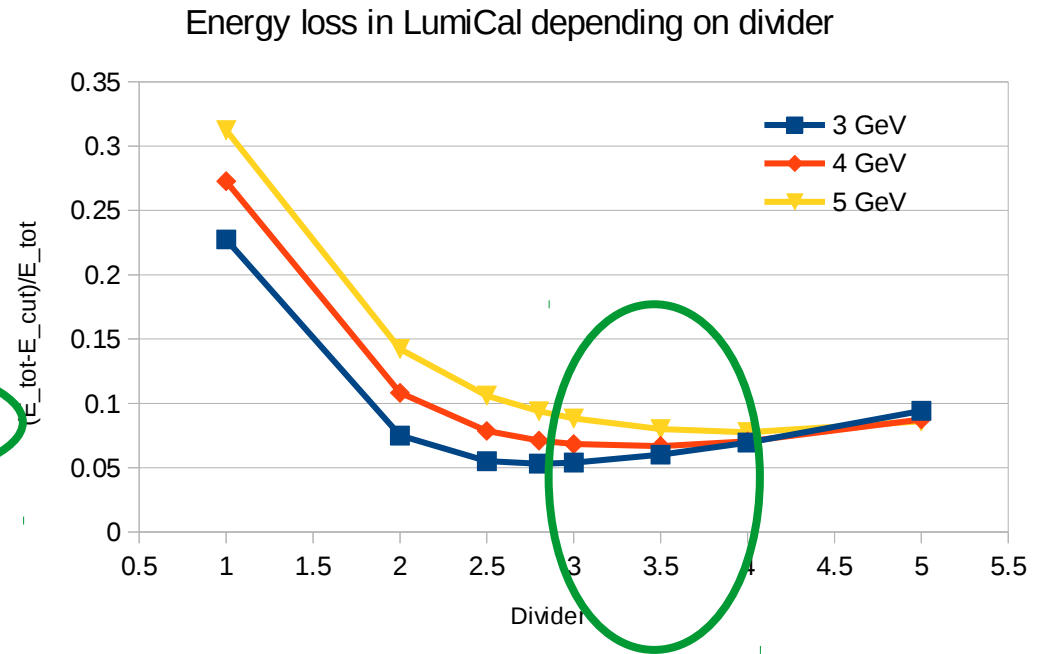
Simulated energy deposition in single sensor pad in 5-th layer (after  $5X_0$ ).





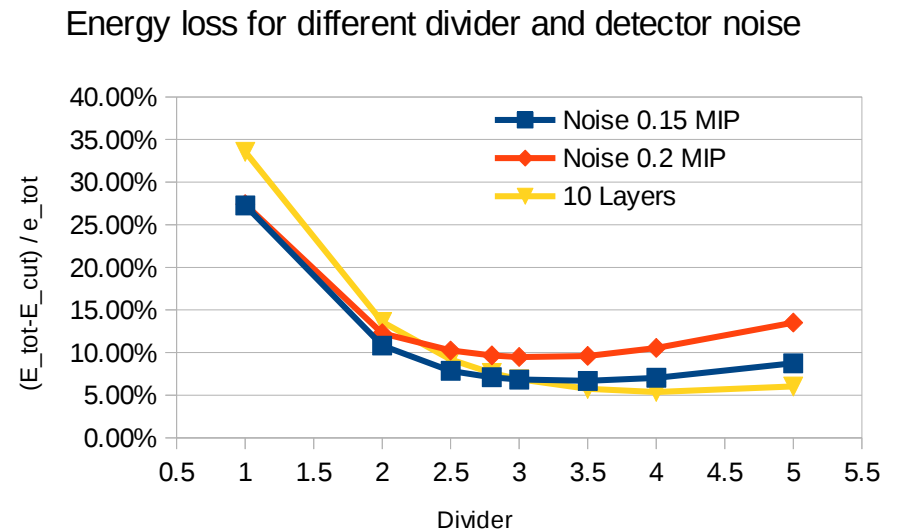
# Charge Divider for Optimizing APV-25 Range

Divider	3 GeV	4 GeV	5 GeV
1	22.75%	27.27%	31.25%
2	7.49%	10.81%	14.23%
2.5	5.50%	7.86%	10.60%
2.8	5.31%	7.10%	9.39%
3	5.39%	6.83%	8.82%
3.5	6.01%	6.68%	8.00%
4	6.95%	7.04%	7.76%
5	9.42%	8.75%	8.59%



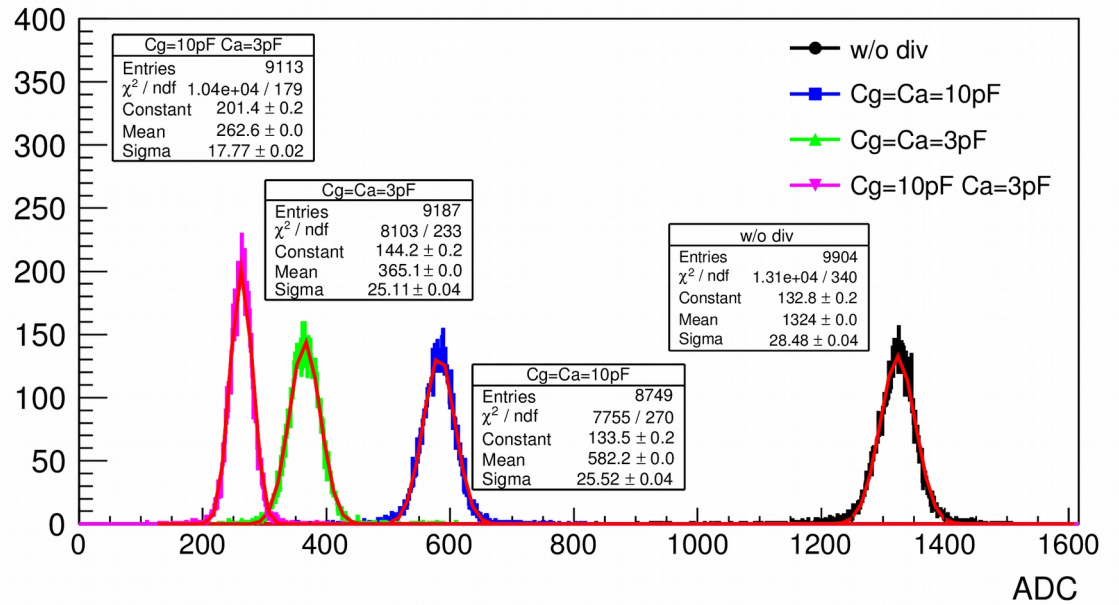
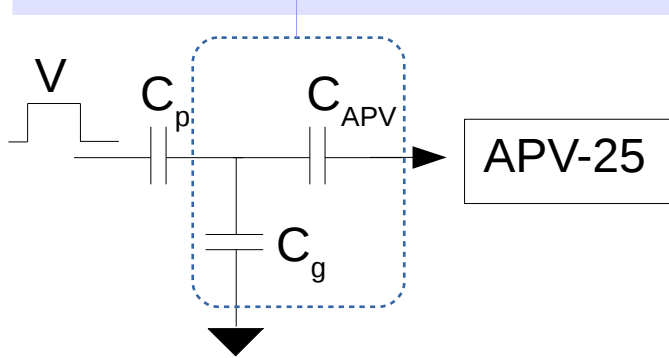
Affect of noise for 4 GeV beam and smaller number of layers (10)

Divider	Noise 0.15 MIP	Noise 0.2 MIP	10 Layers
1	27.27%	27.42%	33.57%
2	10.81%	12.26%	13.60%
2.5	7.86%	10.24%	9.18%
2.8	7.10%	9.67%	7.61%
3	6.83%	9.48%	6.86%
3.5	6.68%	9.60%	5.74%
4	7.04%	10.55%	5.39%
5	8.75%	13.51%	6.03%

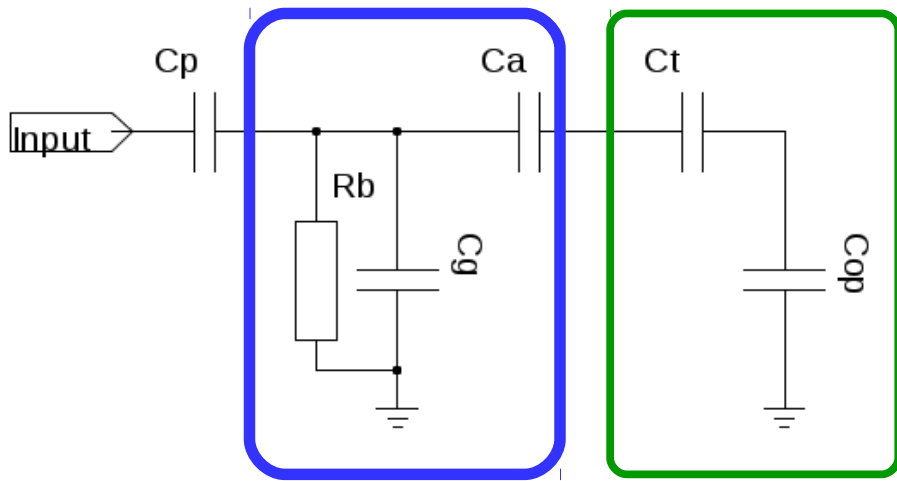


# Divider Implementation Tests

Can be implemented as a small PCB connected to LumiCal detector module and APV-25 board.

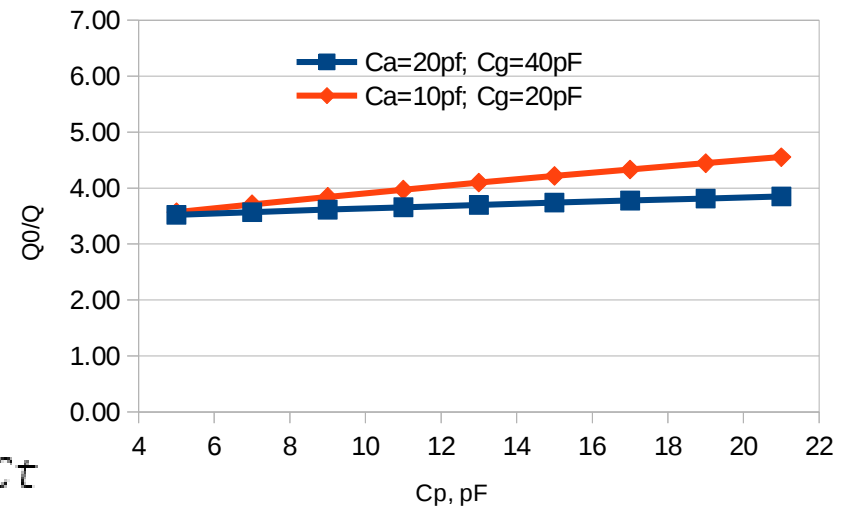


$$C_{op} = C_{fb}(K+1); \quad C_{fb} = 0.15 \text{ pF}; \quad K \sim 10^{10} ?$$



$$(C_g \omega)^{-1} \ll R_b; \quad \frac{Q_0}{Q} = \frac{C_a C_g + C_a C_p + (C_p + C_g + C_a) C_t}{C_a C_t + C_a C_p}$$

Divider vs Input Capacitance



# Identification of Particle Signal with Neural Network

Signal with time response function of CR-RC filter:

$$V = \frac{e^{-\frac{t_0-t}{\tau}} (t-t_0) P_0 \theta}{\tau}$$

$\tau$ , and  $t_0$  are known parameter of the readout

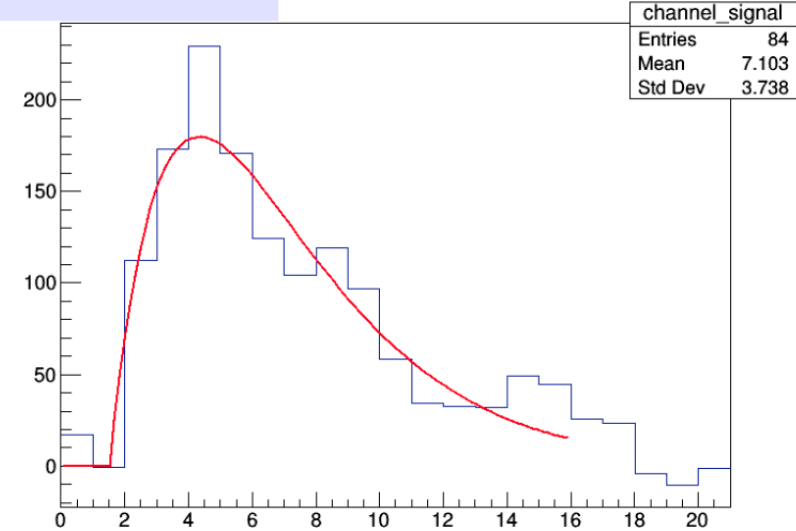
**Neural network is tested to identify the signal in raw data:**

- Input layer 21+1 nodes, one hidden layer with 10+1 node;
- Regularization;
- Training set is generated using the formula with random amplitude and noise generated from Landau and Gaussian distributions respectively.

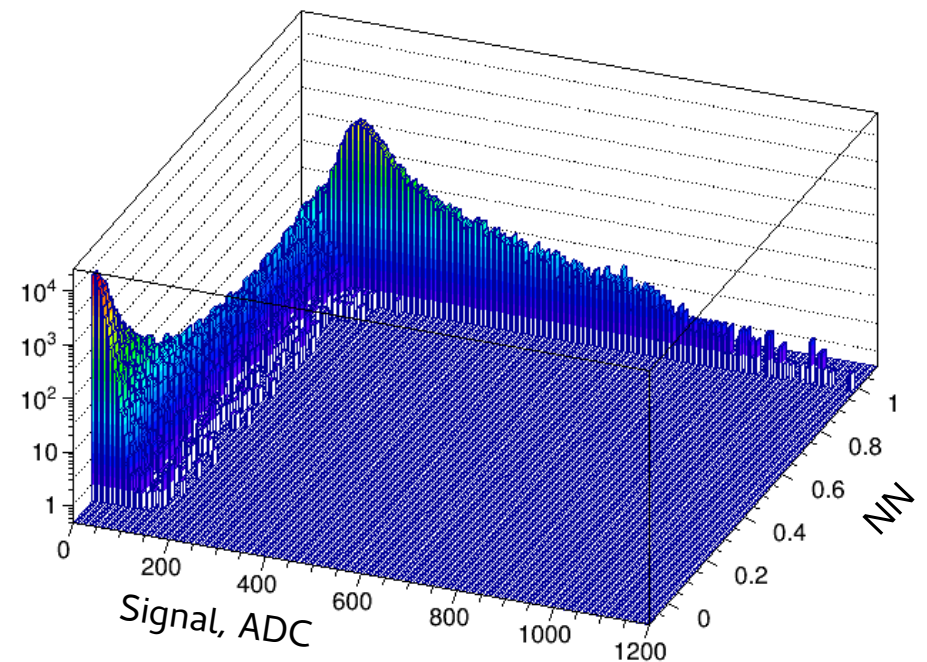
Some specific noise patterns were revealed to be included in training set

NN demonstrate excellent signal identification in noisy environment

Noise  $\sigma=20$  channel\_signal



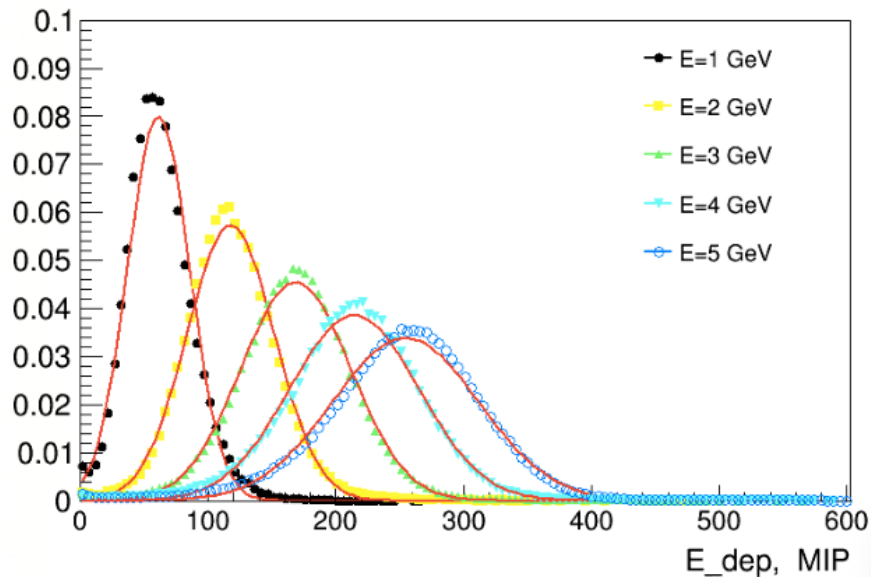
apv\_nn\_output:apv\_signal\_bint1 {apv\_id==2}



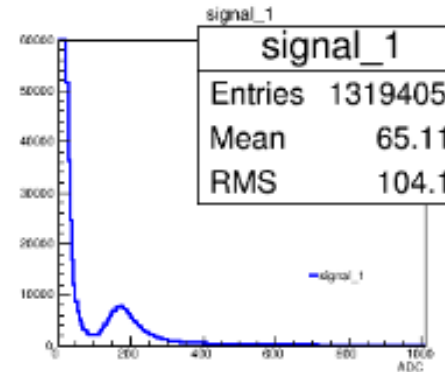
# LumiCal Energy Responce

Cosmic muon events are collected for calibration.

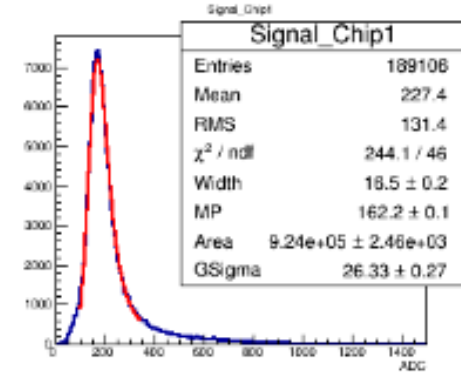
LumiCal response when running with charge divider



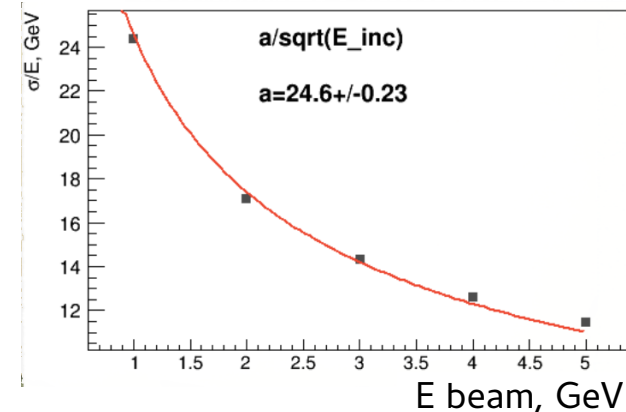
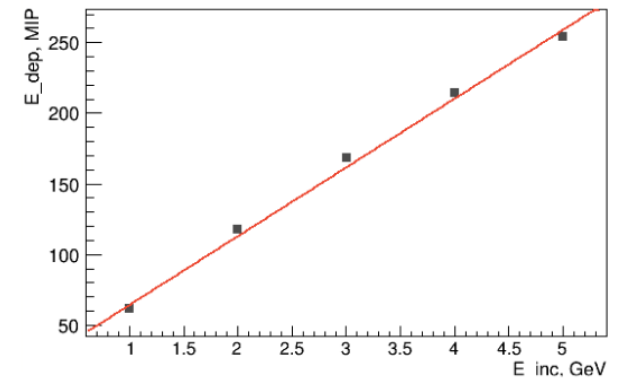
No cuts



NN cut

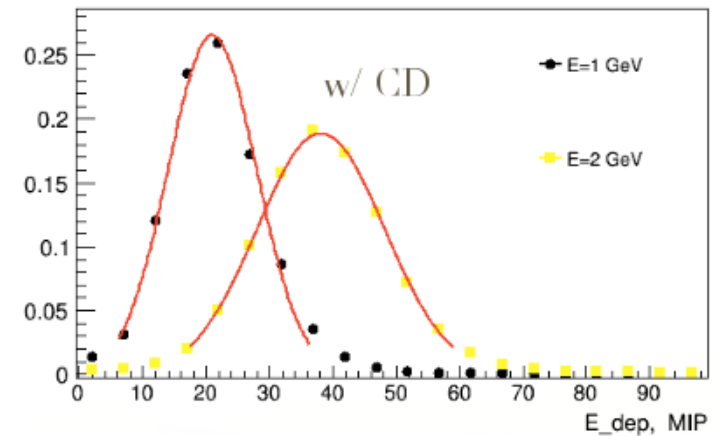


Energy deposited in LumiCal sensor by cosmic muon (ADC)



# Charge Divider Calibration

Calibration factor can be estimated by comparison of the deposited energy in LumiCal with and without charge divider for 1 GeV and 2 GeV beams, where the affect of saturation is relatively small.

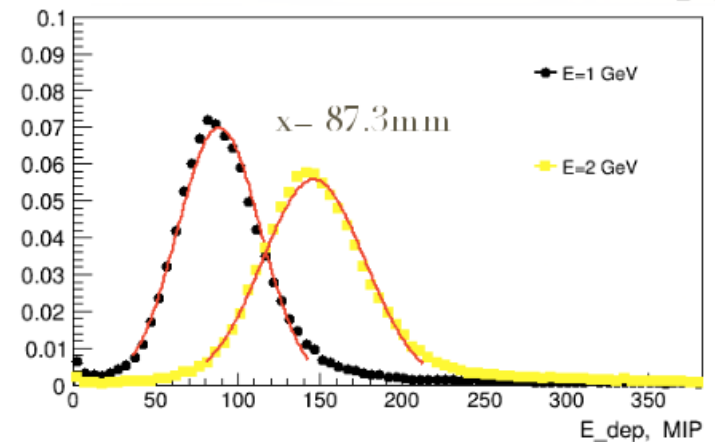


Estimation for

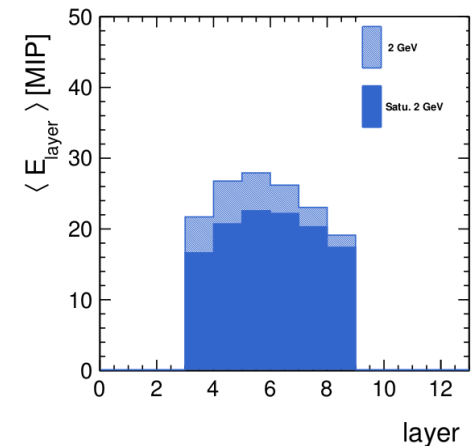
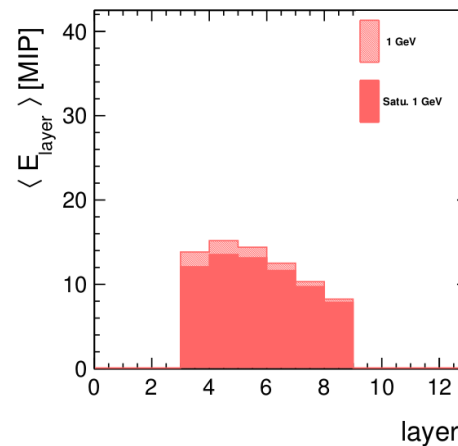
- 1 GeV beam is 4.22 and
- 2 GeV beam – 3.84

The simulations with Geant4 showed that correction for the saturation is:

- 10 % for 1 GeV beam and
- 20 % for 2 GeV beam

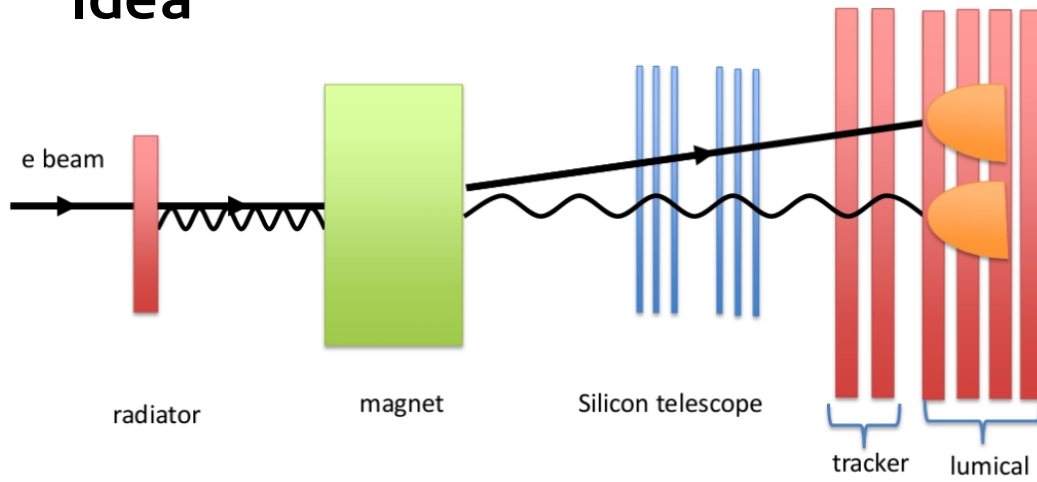


- Taking into account this correction the calibration factor in both measurements is around 4.7.
- As expected there is dependence on the input capacitance observed in different beam positions.



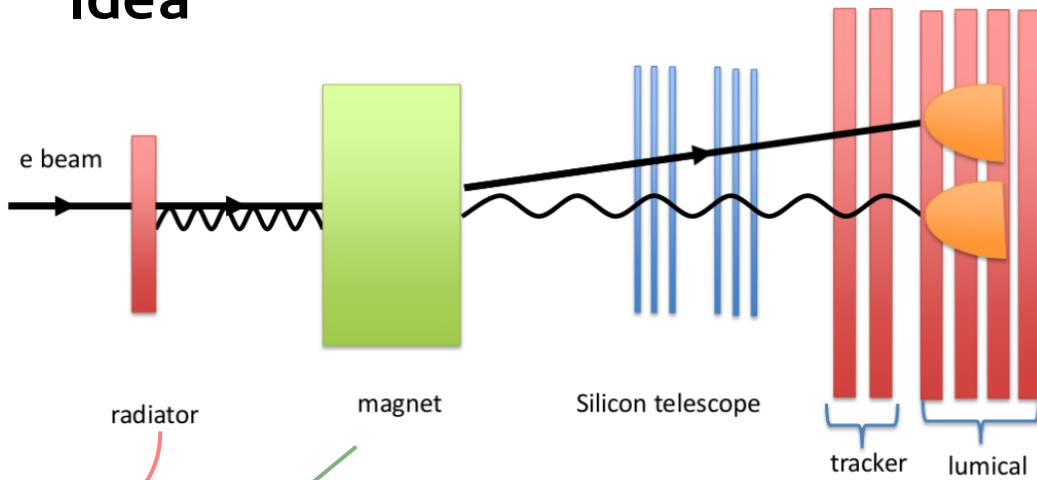
# Test-beam Design

## Idea

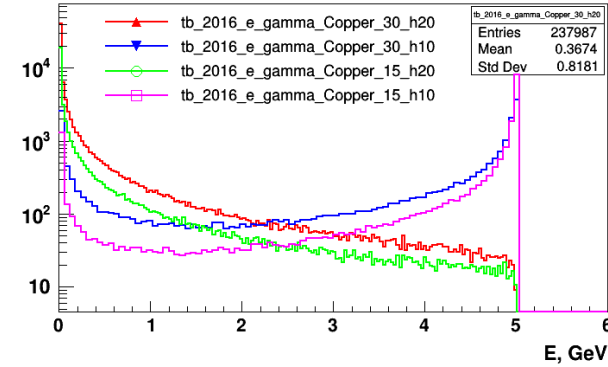


# Test-beam Design

## Idea



Electrons

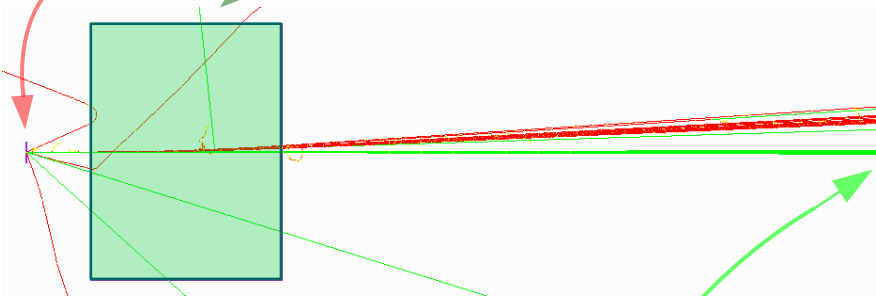


e<sup>-</sup> of 5 GeV

1.5 mm of Copper target

Number of secondaries per event :

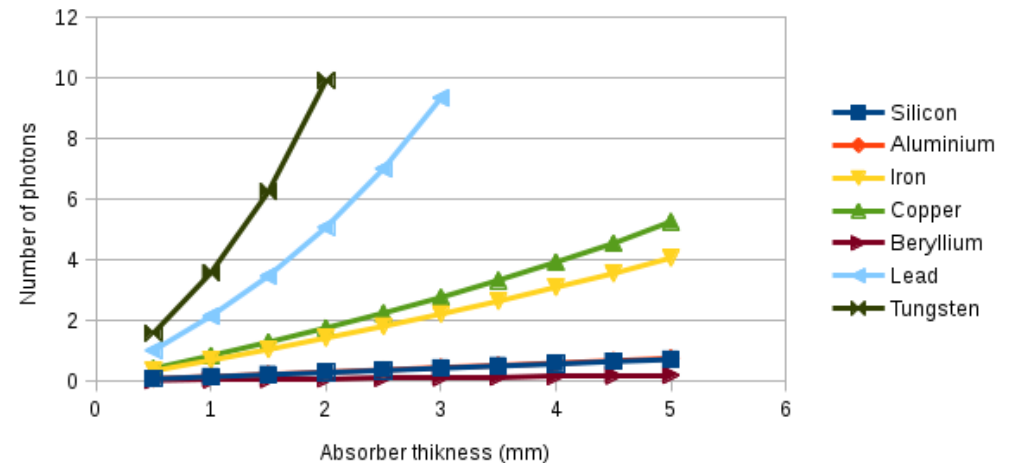
- Gammas = 1.28;
- electrons = 0.5193;
- positrons = 0.02402



Geant4 simulation

Photons

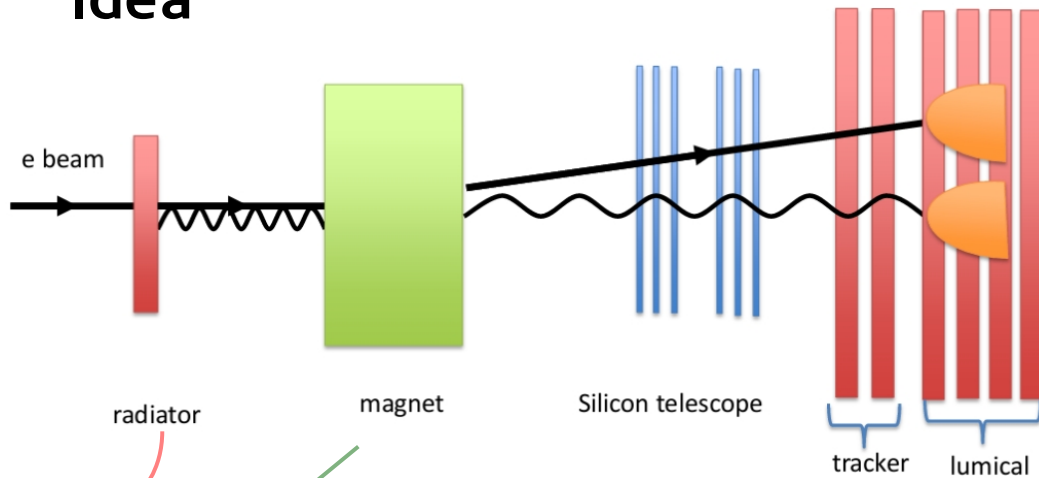
Number of secondary photons from 5 GeV electron





# Test-beam Design

## Idea

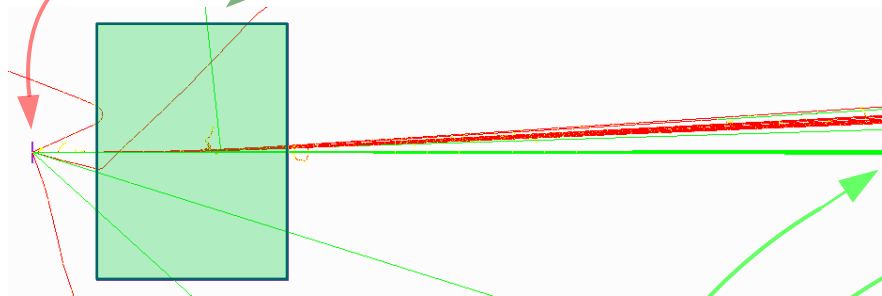


Electrons

Need to optimize

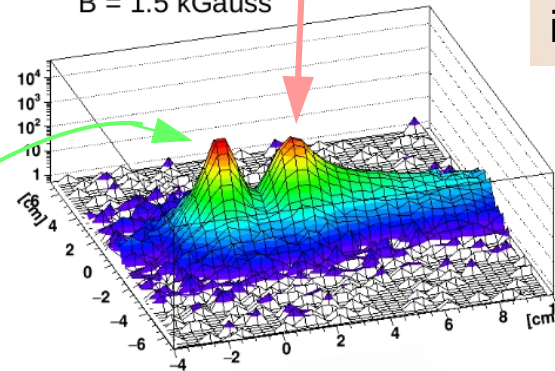
- B field,
- positions of telescope and LumiCal

so that both e and  $\gamma$  beams go through telescope and can be spatially resolvable in LumiCal.

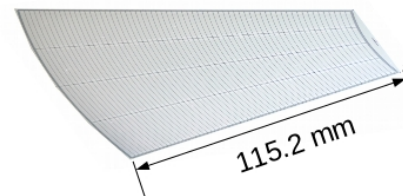


Geant4 simulation

B = 1.5 kGauss

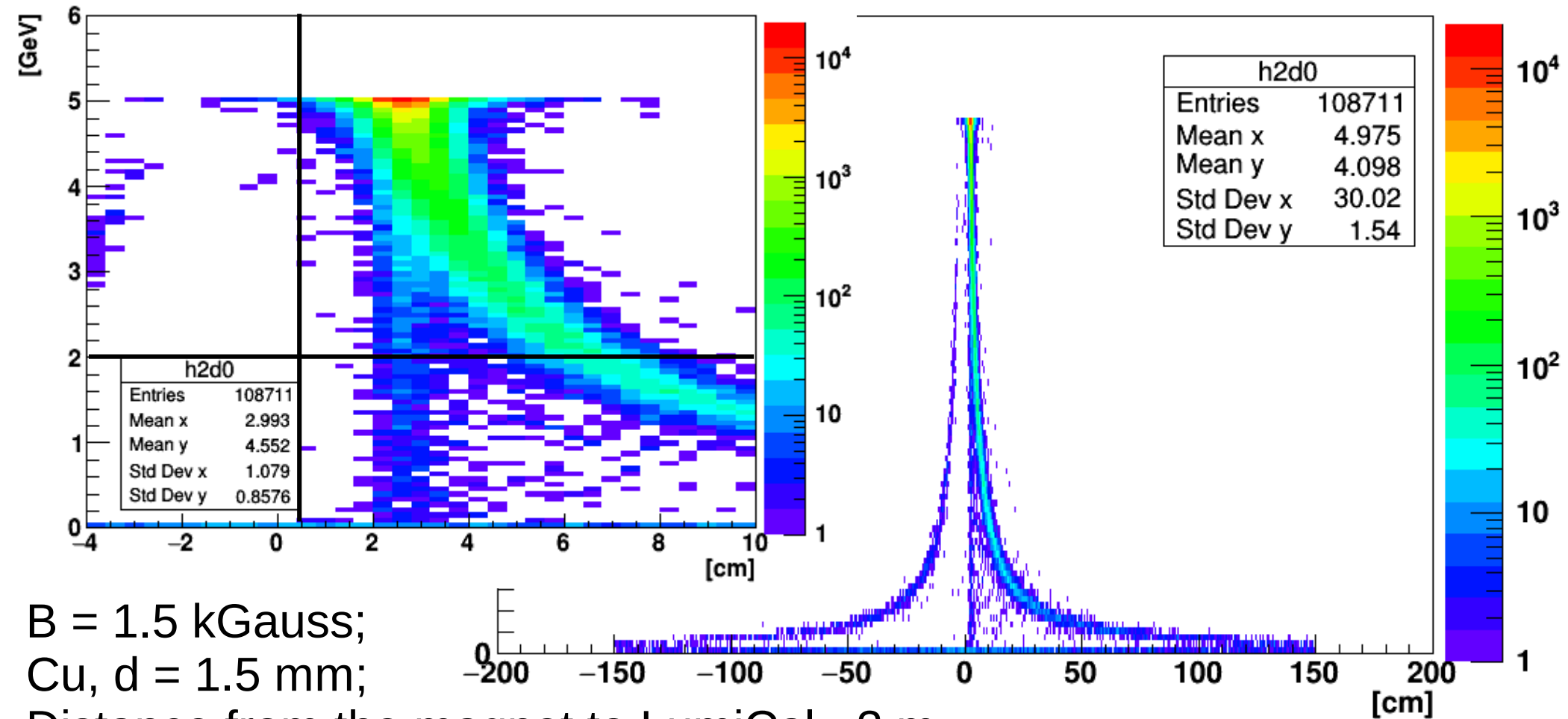


Photons



# Electron Position vs Energy

(transmit, charged) : projected position at exit vs energy



B = 1.5 kGauss;

Cu, d = 1.5 mm;

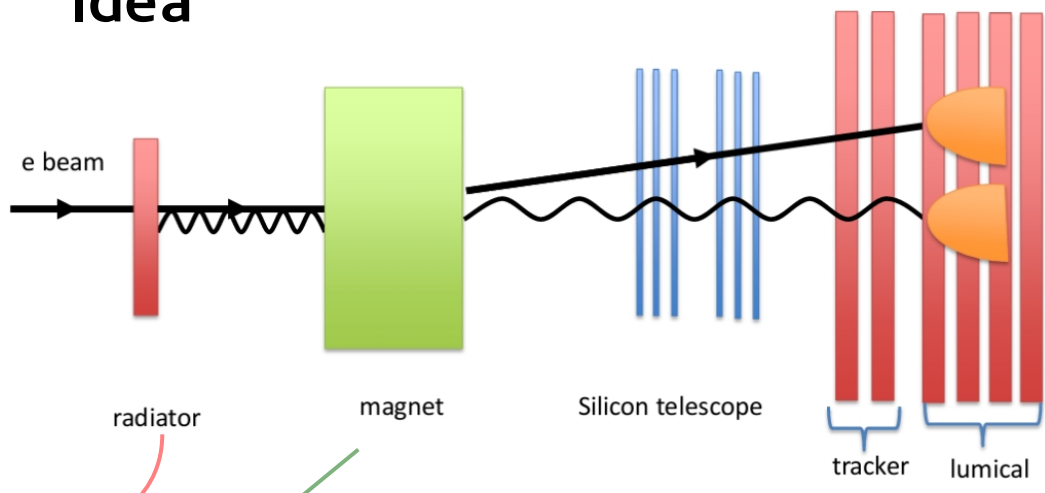
Distance from the magnet to LumiCal ~3 m.

Electron energy in the range 2 GeV – 5 GeV : 86% of events  
 3 GeV – 5 GeV : 82% of events

2 GeV – 4 GeV : 10%  
 3 GeV – 4 GeV : 6.6%

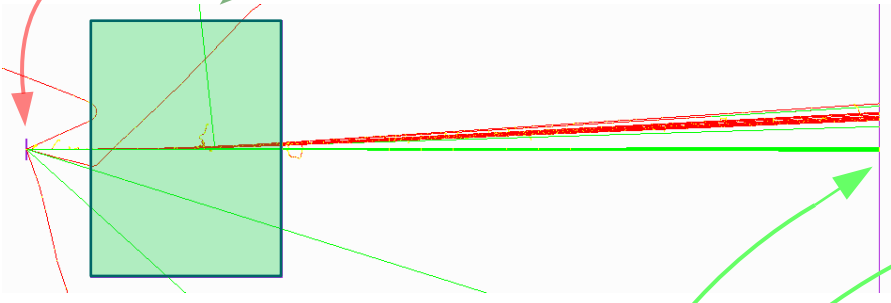
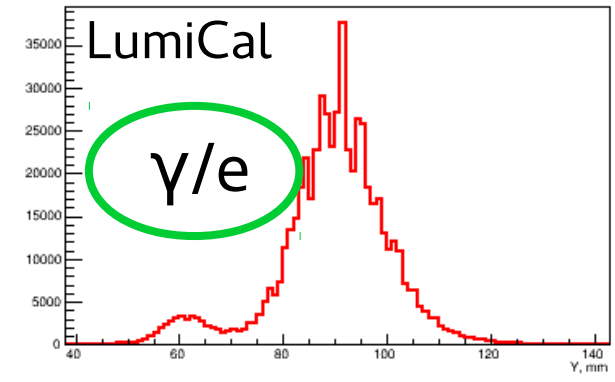
# Test-beam Design

## Idea

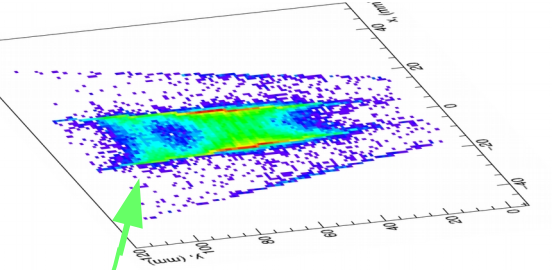
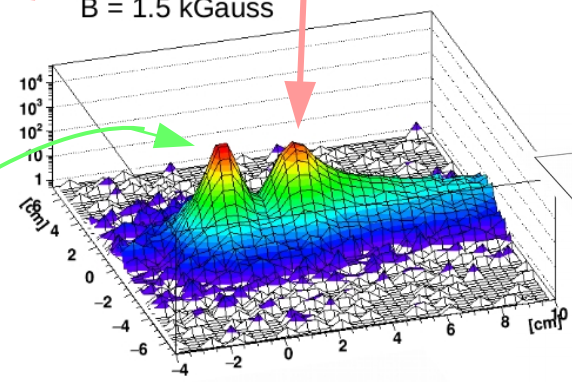


Electrons

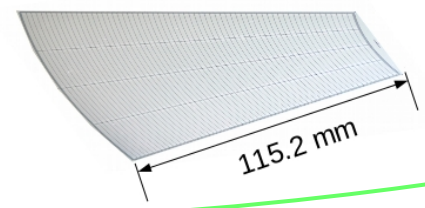
Observed occupancy in



B = 1.5 kGauss



Photons



# Clustering Algorithms

## E-clustering:

First looks for hits with local maximum of deposited energy and consider them as seeds of the clusters.

Then all neighboring pads with descending energy are assigned to the seeds. Used in Zeus:

- Capable of resolving spatially joined clusters;
- Sensitive to the fluctuation in shower development.

## Linking neighboring pads:

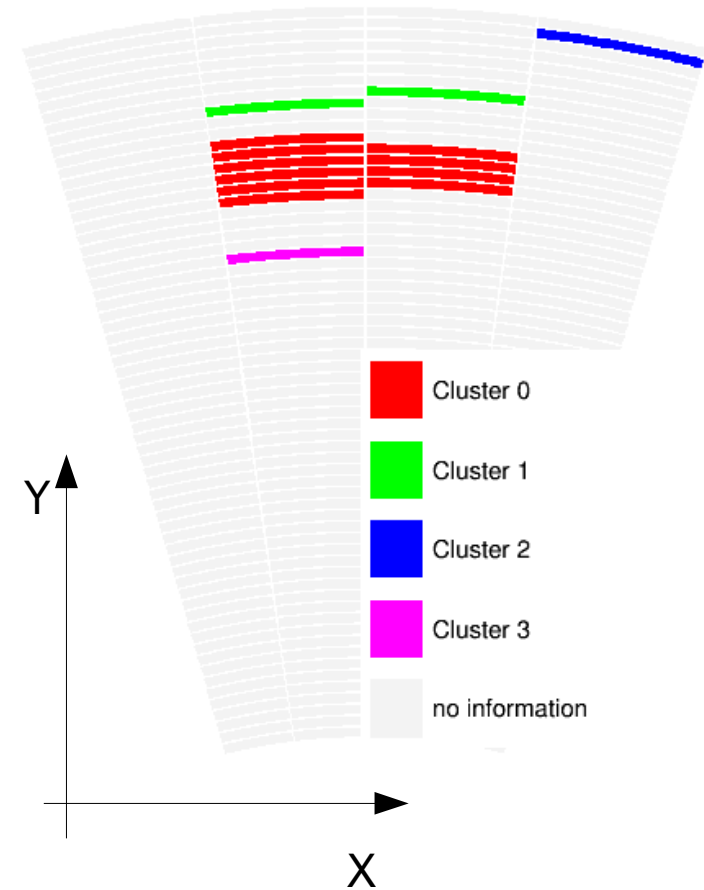
Looks for the closest neighbors (with distance no more than 1 pad in any direction) and then collects them to the cluster

- Very simple.

## k-means:

Widely used in machine learning. Essentially, assigns points to cluster centers and locate them to minimize sum of the distances:

- Different implementations are available, easy to try;
- It assumes a certain given number of clusters and does not use all physics information;



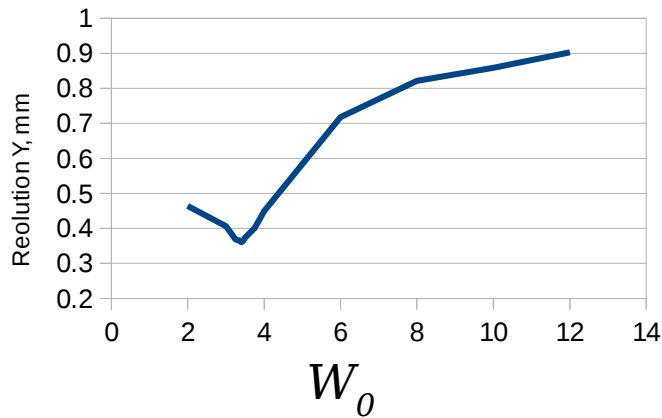
# Cluster Position Reconstruction in Simulations

- Logarithmic weighting:

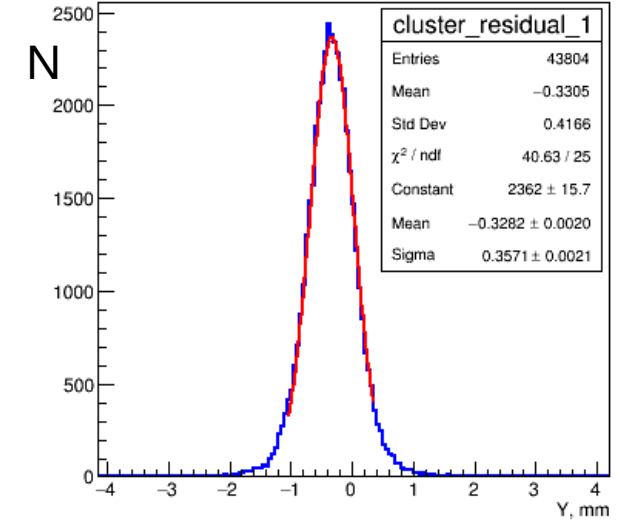
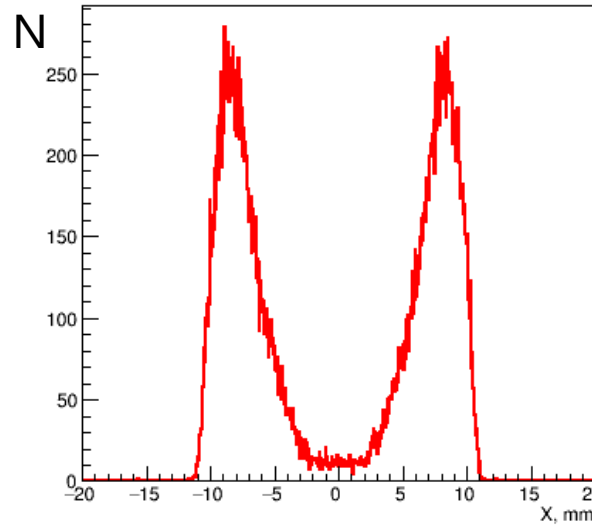
$$Y_s = \frac{\sum_n n w_n}{\sum_n w_n},$$

$$w_n = \max \left\{ 0; W_0 + \ln \frac{E_n}{\sum_n E_n} \right\},$$

Logarithmic Weighting Constant



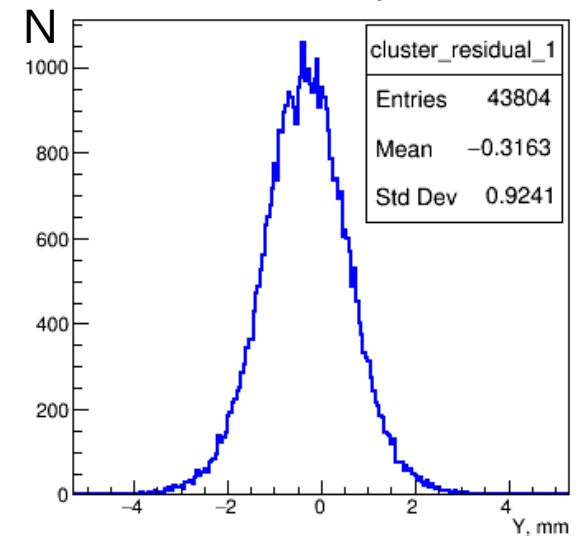
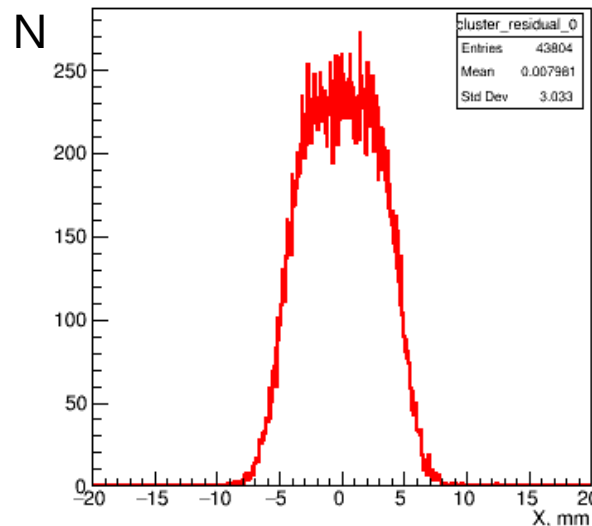
At  $W_0 = 3.4$   
Y resolution is 0.36 mm



$W_0 = 3.4$

Distributions of X and Y residuals for

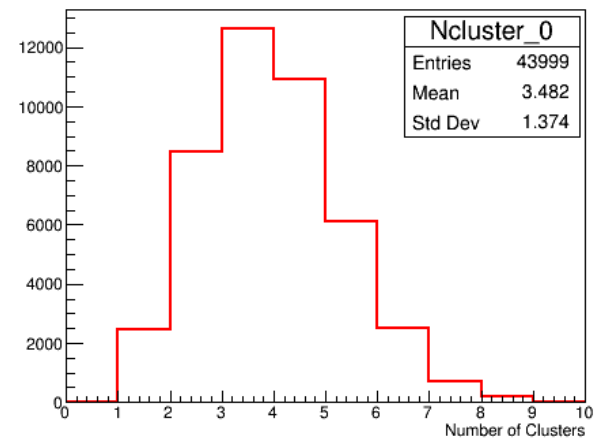
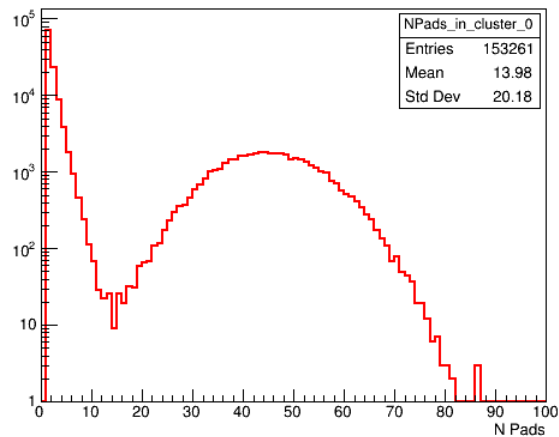
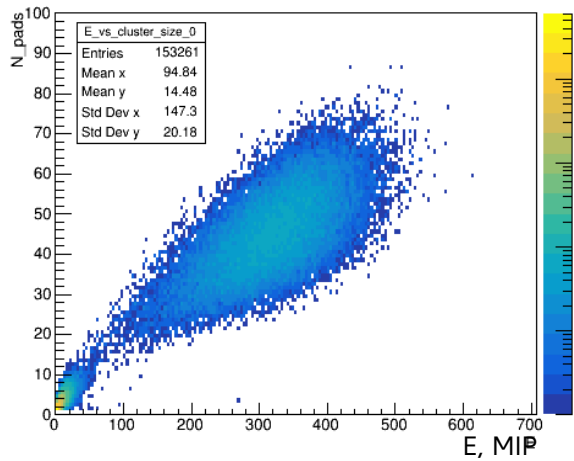
$W_0 = 12.0$



# Clusters: MC vs Data

5 GeV electron beam

MC Simulations

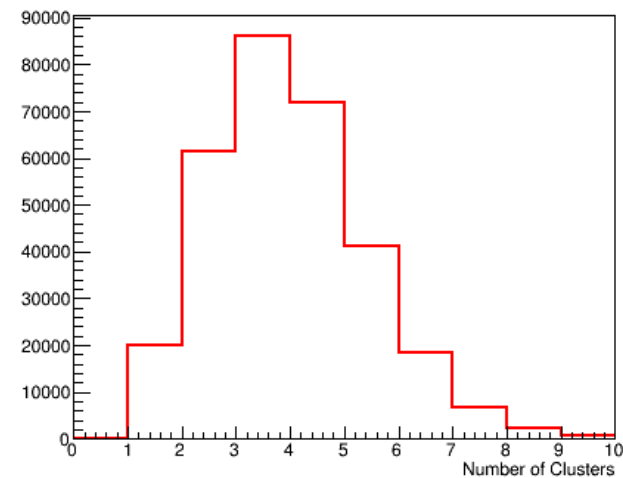
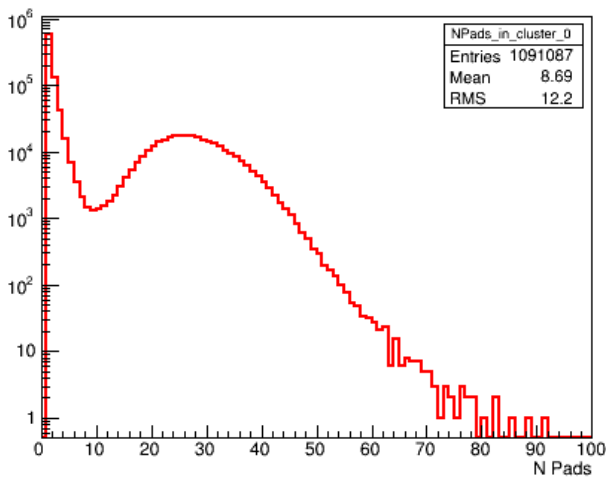
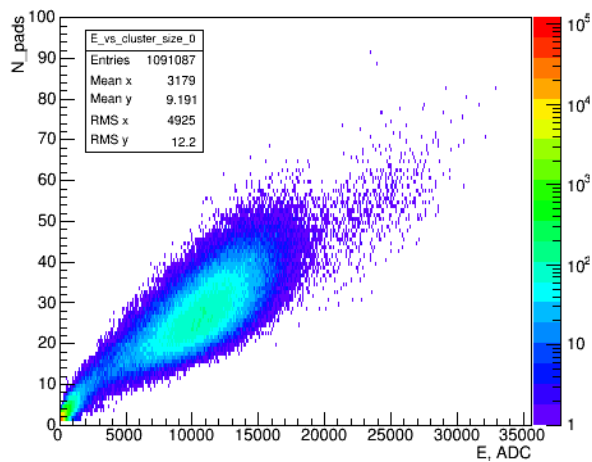


Cluster energy vs size

Cluster size

Number of clusters

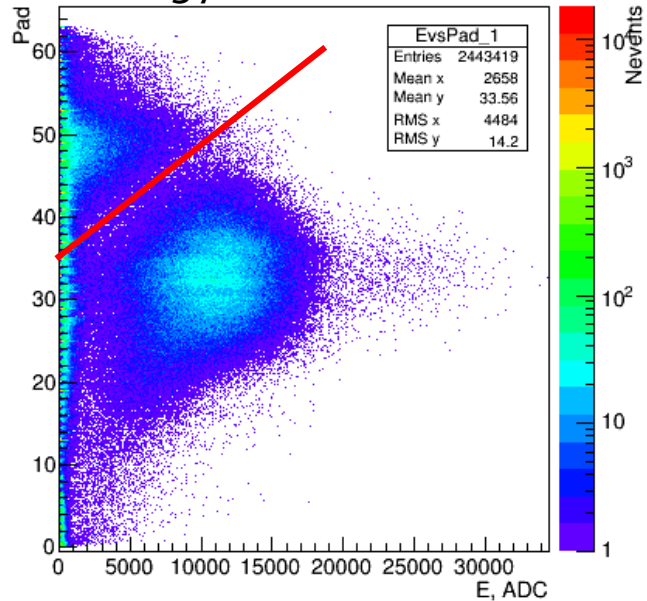
Data



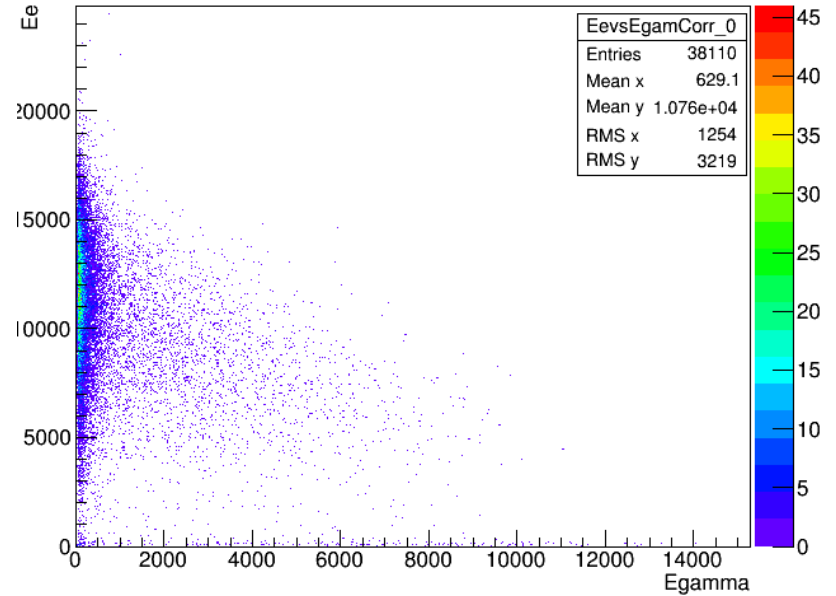
Simulation does not take into account electronic noise, which might explain the difference in cluster size

# Electron and Photon Energy

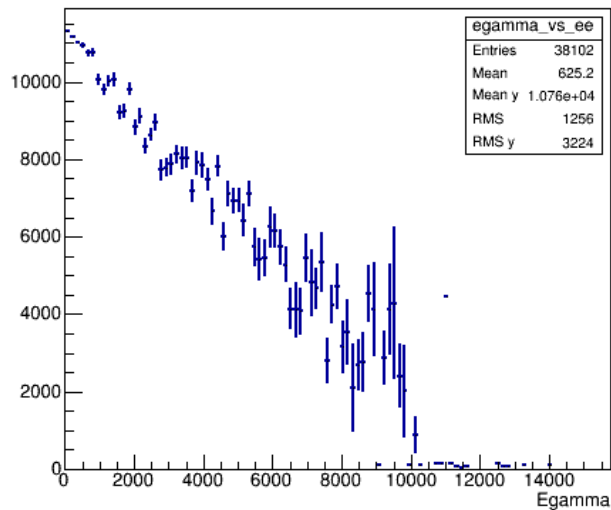
Clusters position vs energy. Data.



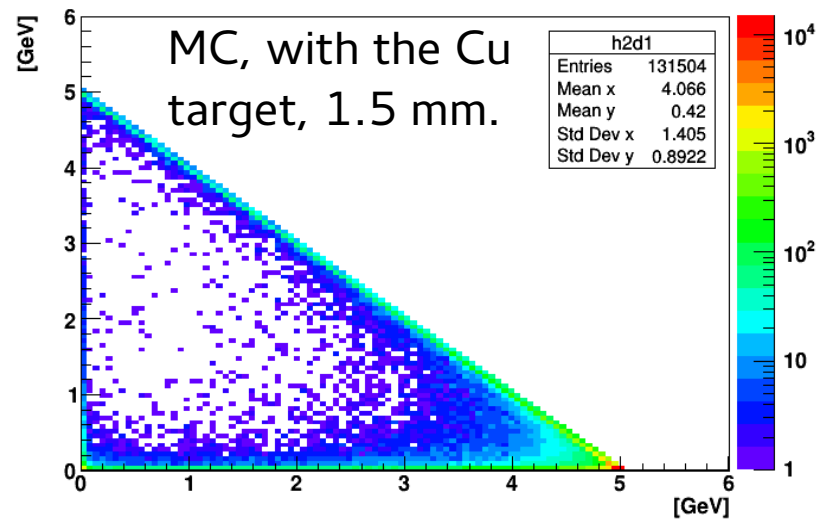
$E_e$ ,  $E_\gamma$  correlation, Data



$E_e$  vs  $E_\gamma$  profiling plot. Data.



(transmit, charged vs neutral): kinetic energy e-gamm at exit





# Summary

- The design of the beam test setup and detector components were studied and optimized in simulations. It has been successfully realized at DESY and allowed to collect data for e/ $\gamma$  identification study in LumiCal combined with tracking detector.
- Neural network was successfully applied to identify the particle signal in raw detector data.
- Different cluster reconstruction algorithms were implemented and tested in simulation and data. Their performance is in good agreement for MC and data.
- Cluster position reconstruction with logarithmic weighting algorithm was optimized in simulation ( $W_0=3.4$ ). It will be compared to the data using the position of the particle reconstructed in telescope.
- Ongoing study on back scattering and e/ $\gamma$  identification.