

Tangerine: Monte Carlo simulations of MAPS in a 65 nm imaging process

Manuel Alejandro Del Rio Viera on behalf of the Tangerine collaboration

10th Beam Telescopes and Test Beams Workshop

Lecce, June 21st 2022



The Tangerine Project - Towards Next Generation Silicon Detectors

Goal: Design a monolithic pixel detector in 65nm CMOS imaging technology

- **Main application:** beam telescope
- **Potential applications:** linear collider experiments, etc.

Performance targets:

- Position resolution $\leq 3 \mu\text{m}$
- Time resolution $\sim 1 - 10 \text{ ns}$
- Material budget $\sim 50 \mu\text{m Si}$



MIMOSA Beam Telescope, DESY II Test Beam Facility

The Tangerine Project - Towards Next Generation Silicon Detectors

- **The Tangerine project's goal** is to develop the next generation of small collection electrode **monolithic silicon pixel detectors** using the **65 nm CMOS** imaging process.



Monolithic Silicon Pixel Detector using 65 nm CMOS imaging process:

- **Higher logic density**
- **Lower power consumption** (compared to previously used processes).

In monolithic sensors:

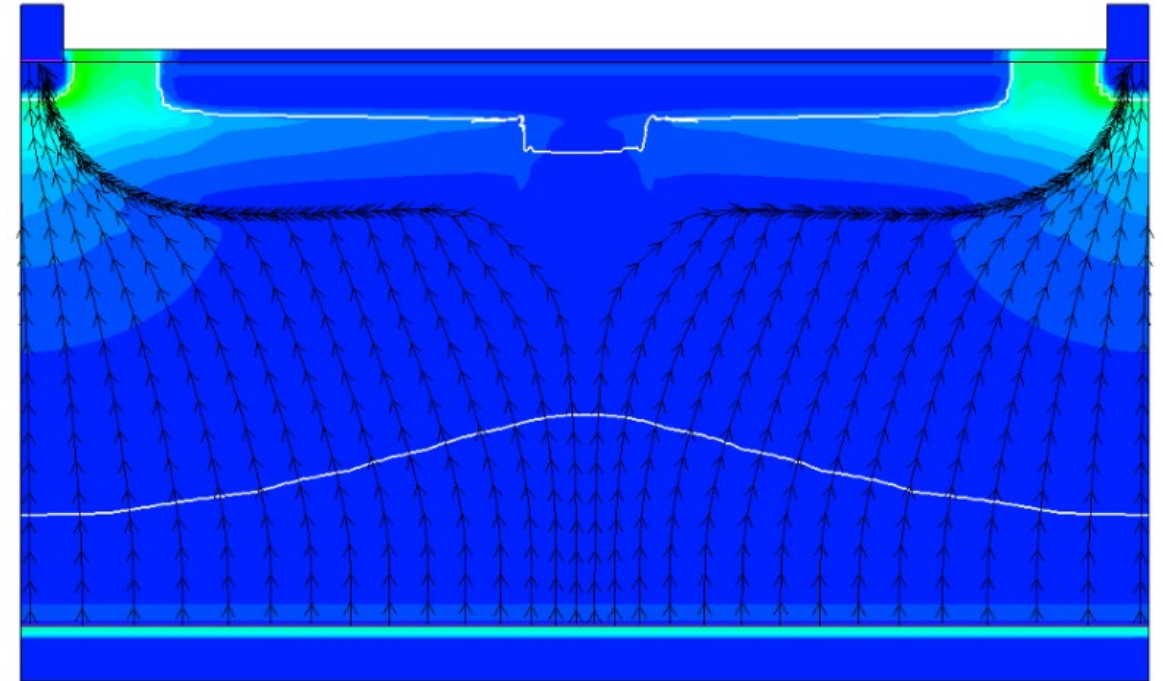
- The **sensitive volume and readout are in a single chip**
- **Lower material budget, and reduced cost and production effort** (compared to hybrid sensors).

Technology computer-aided design (TCAD)

The **electric field** and the **doping concentration** in silicon detectors are certainly **complicated**.

TCAD is utilized in order to obtain this important profiles that characterize the **detector layout**.

SYNOPSYS®



<https://doi.org/10.48550/arXiv.2206.05474>

Shoutout to **Anastasiia Velyka, Adriana Simancas, Larissa Mendes**



Allpix Squared (Allpix²)

A combination of **TCAD** and **Monte Carlo (MC)** simulations are used.

Allpix² is a generic **simulation framework for silicon detectors**, written in modern C++.



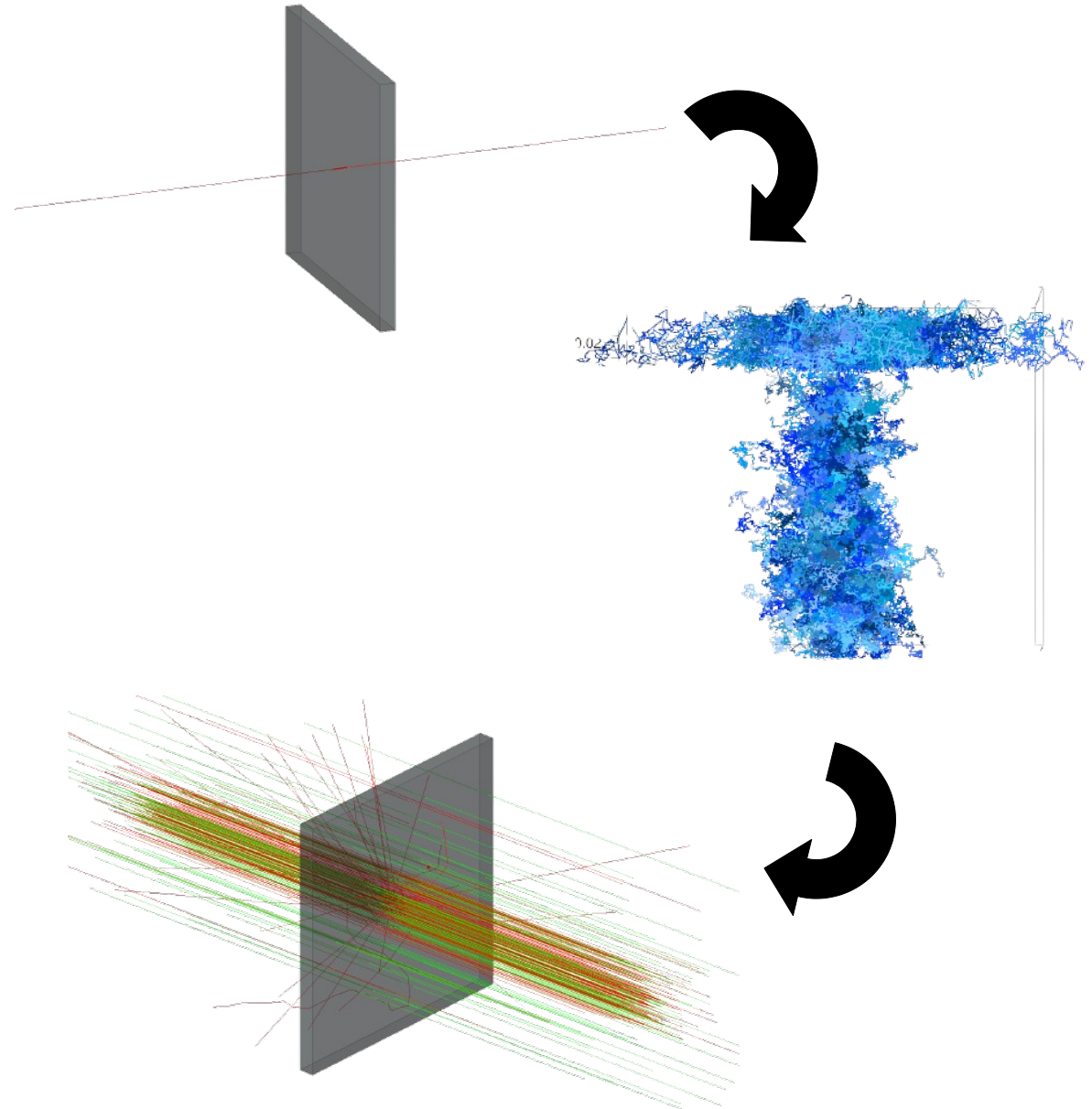
Motivation:

- Reduced simulation **time** and **simplicity** compared to the **TCAD** only approach
- **Value** of the simulation results
- Reduced **cost**
- **Reproducibility** of performance

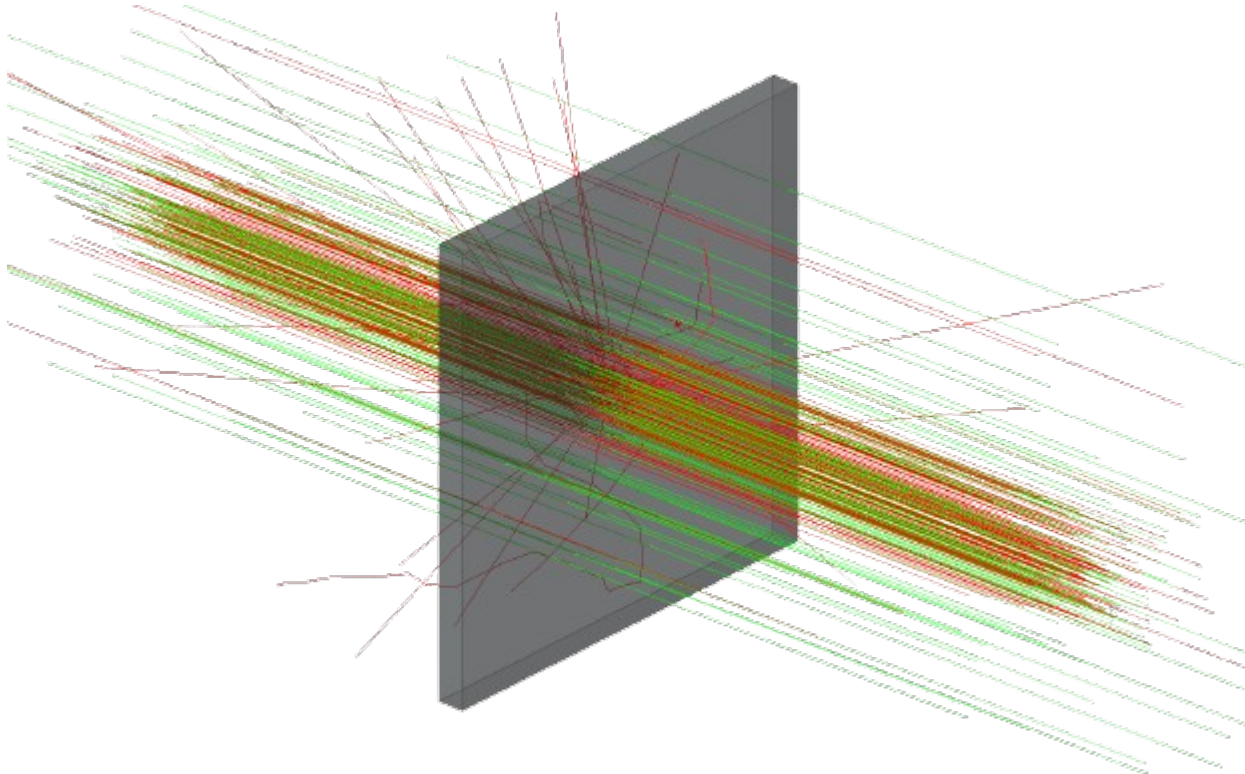
Monte Carlo Setup

- 1) A particle is **randomly*** shoot through the sensor.
- 2) **Ionization** (And other Physics processes: Diffusion, Drift, Recombination...).
- 3) **Repetition** (Same energy and direction).
- 4) **Analysis**

* Gaussian Beam like



Monte Carlo Setup



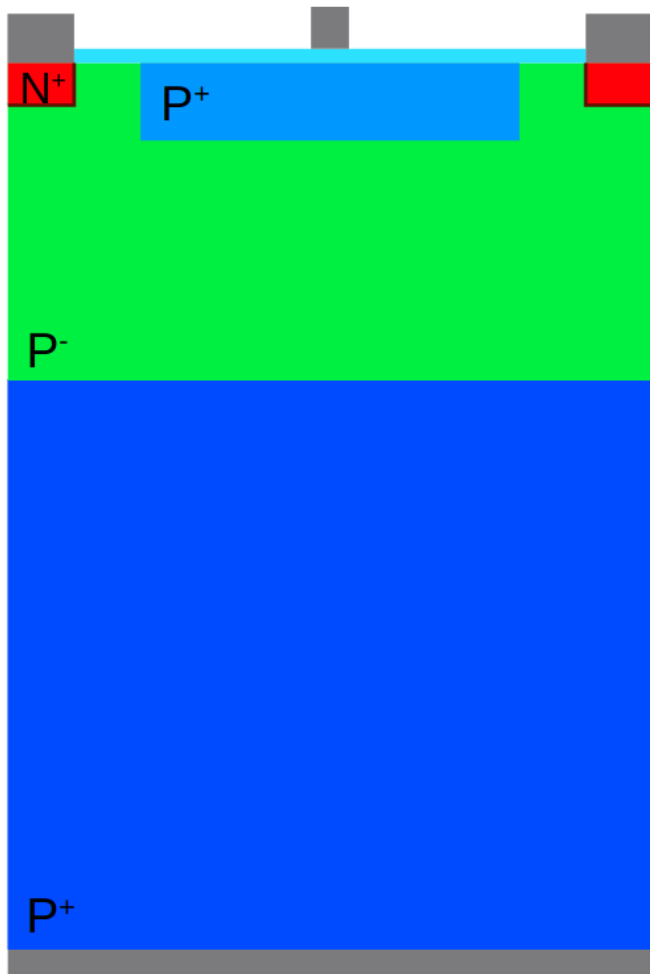
In our case, the beam is a Gaussian electron beam with a **5 GeV** energy (a **DESY** like type of beam).

The number and size of the pixels in the sensor can be adjusted depending in our needs, a typical size being **20x20 μm^2** .

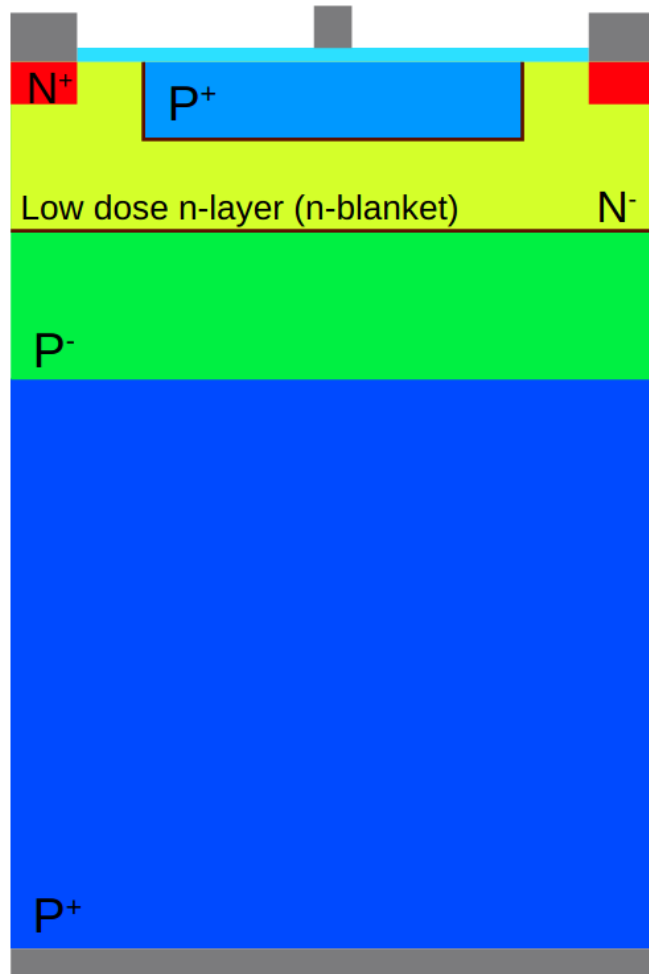
Important quantities for us to obtain are:

- **Efficiency**
- **Hit Map**
- **Cluster Size and Cluster Charge**
- **Spatial Resolution**
- ...

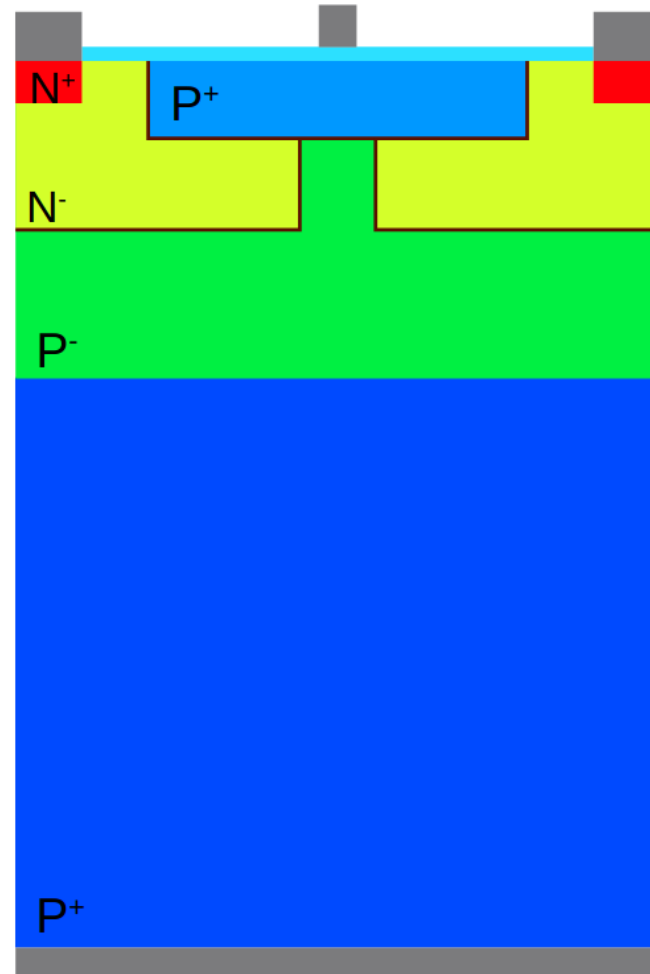
Standard Layout

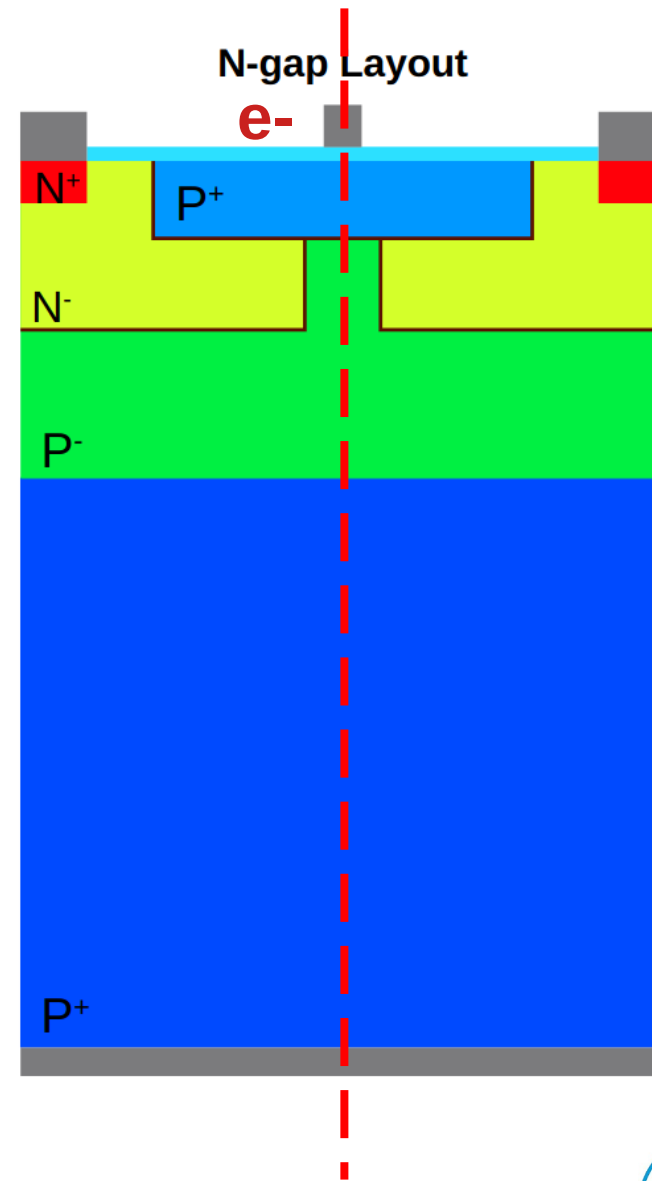
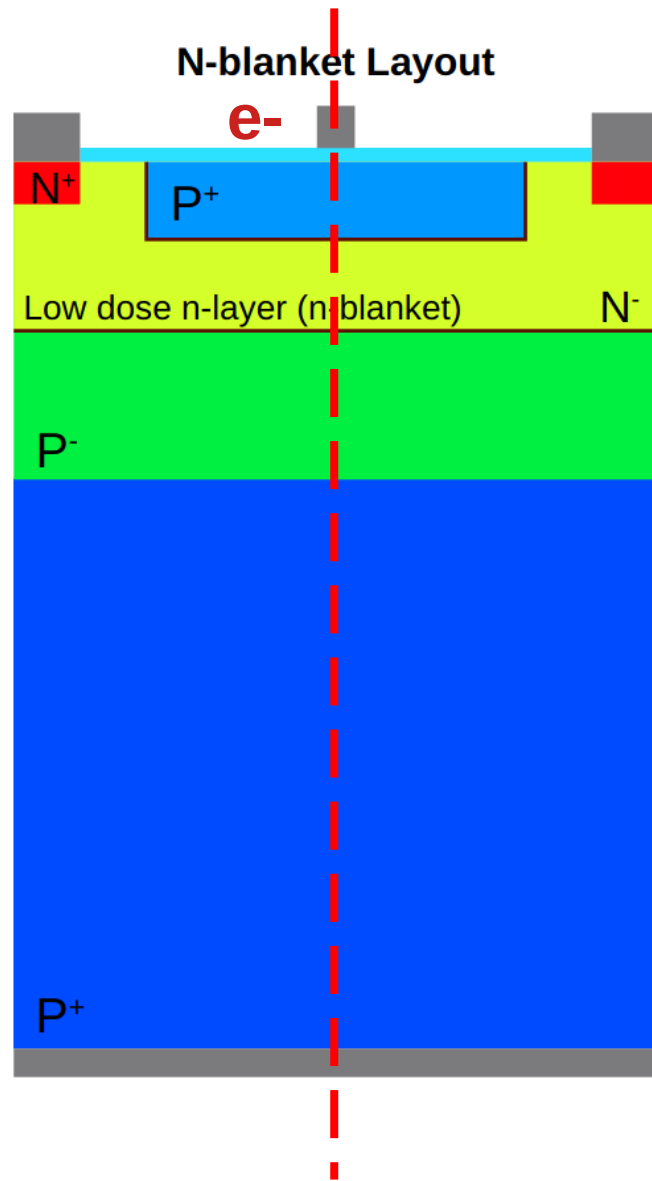
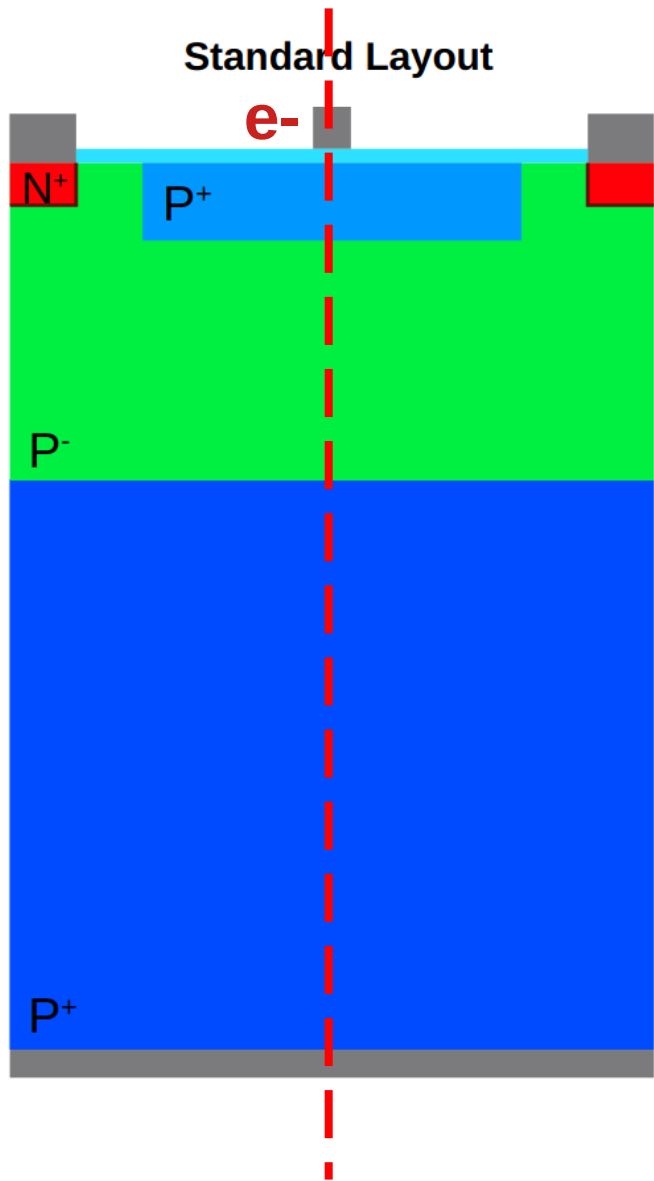


N-blanket Layout

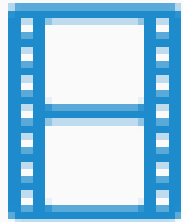


N-gap Layout

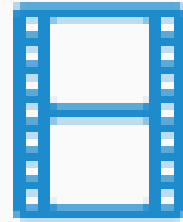




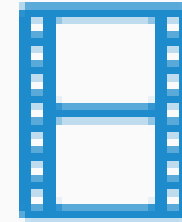
Standard layout



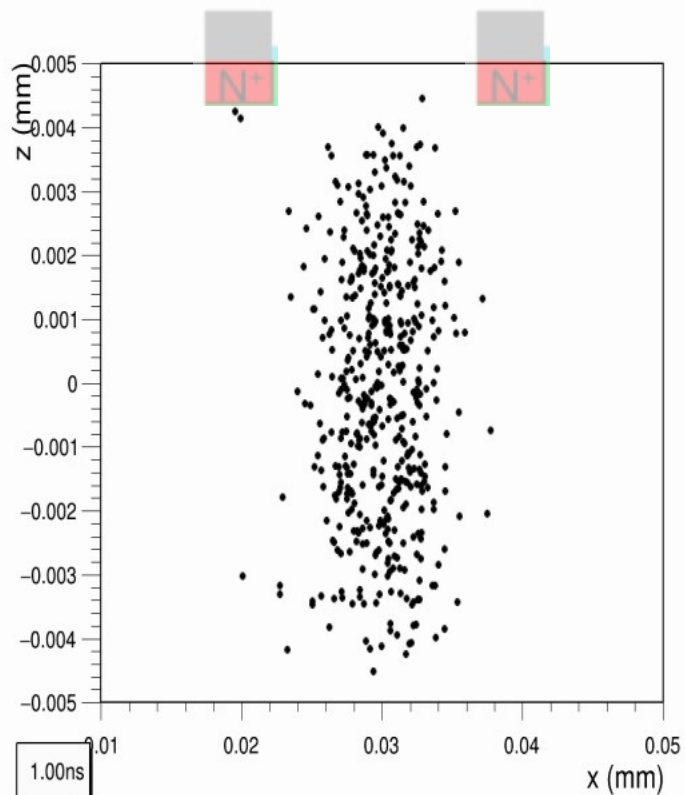
N-blanket layout



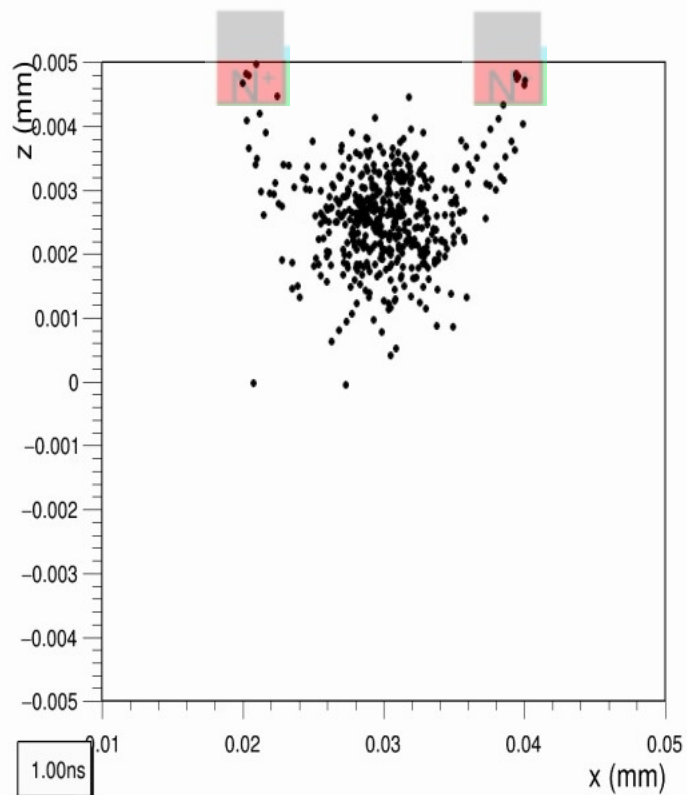
N-gap layout



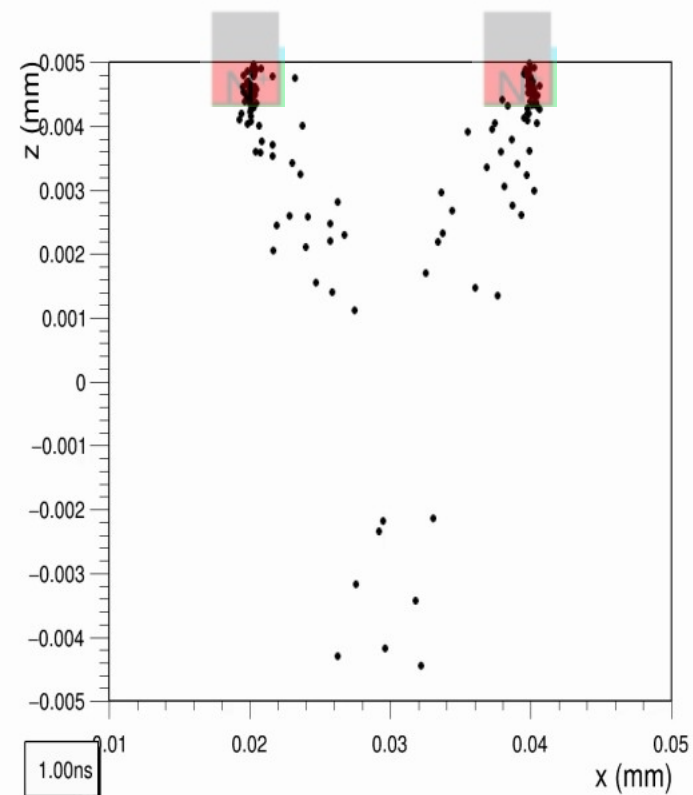
Standard layout



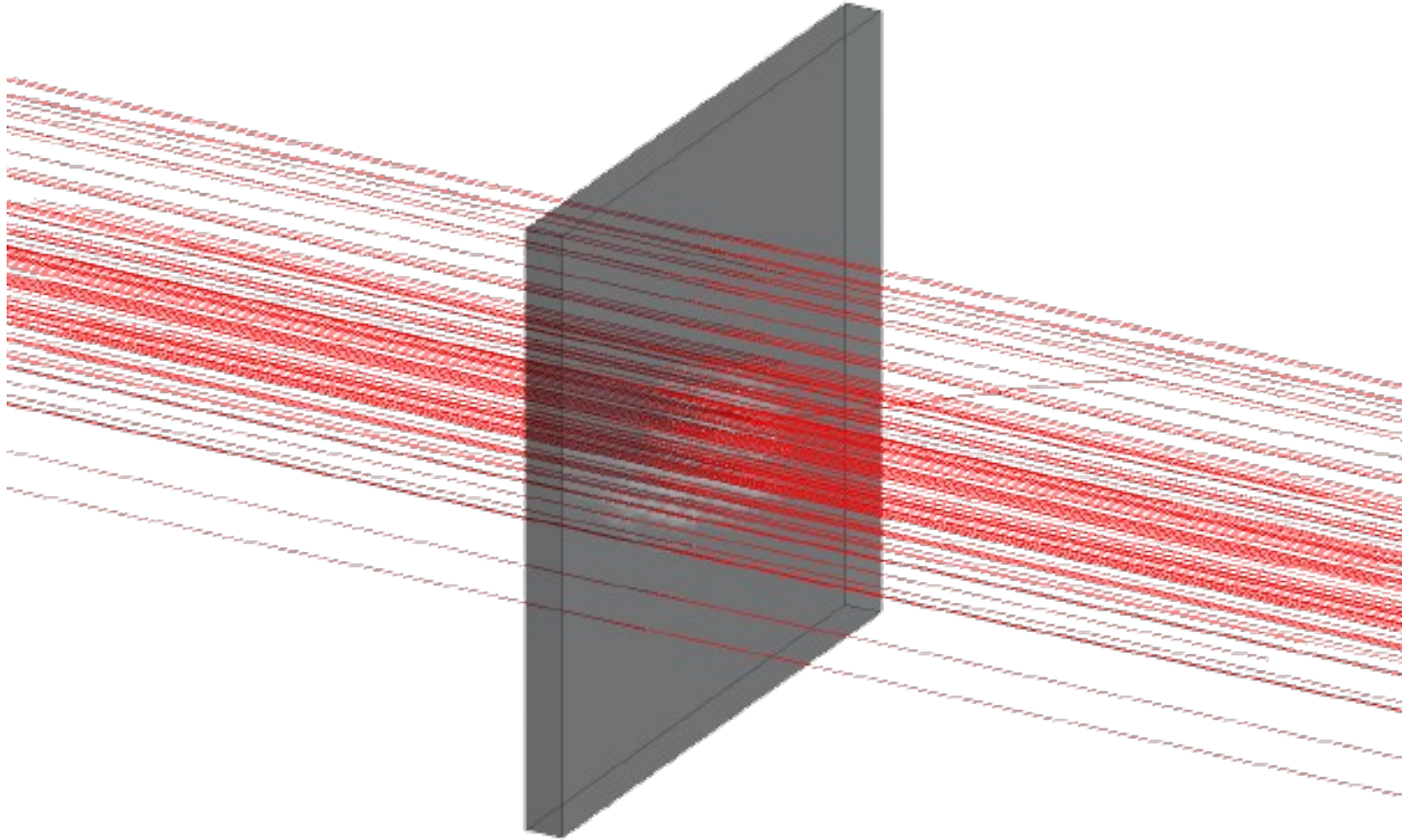
N-blanket layout



N-gap layout



Efficiency



$$\text{Efficiency} = \frac{\text{Fired}}{\text{Events}}$$

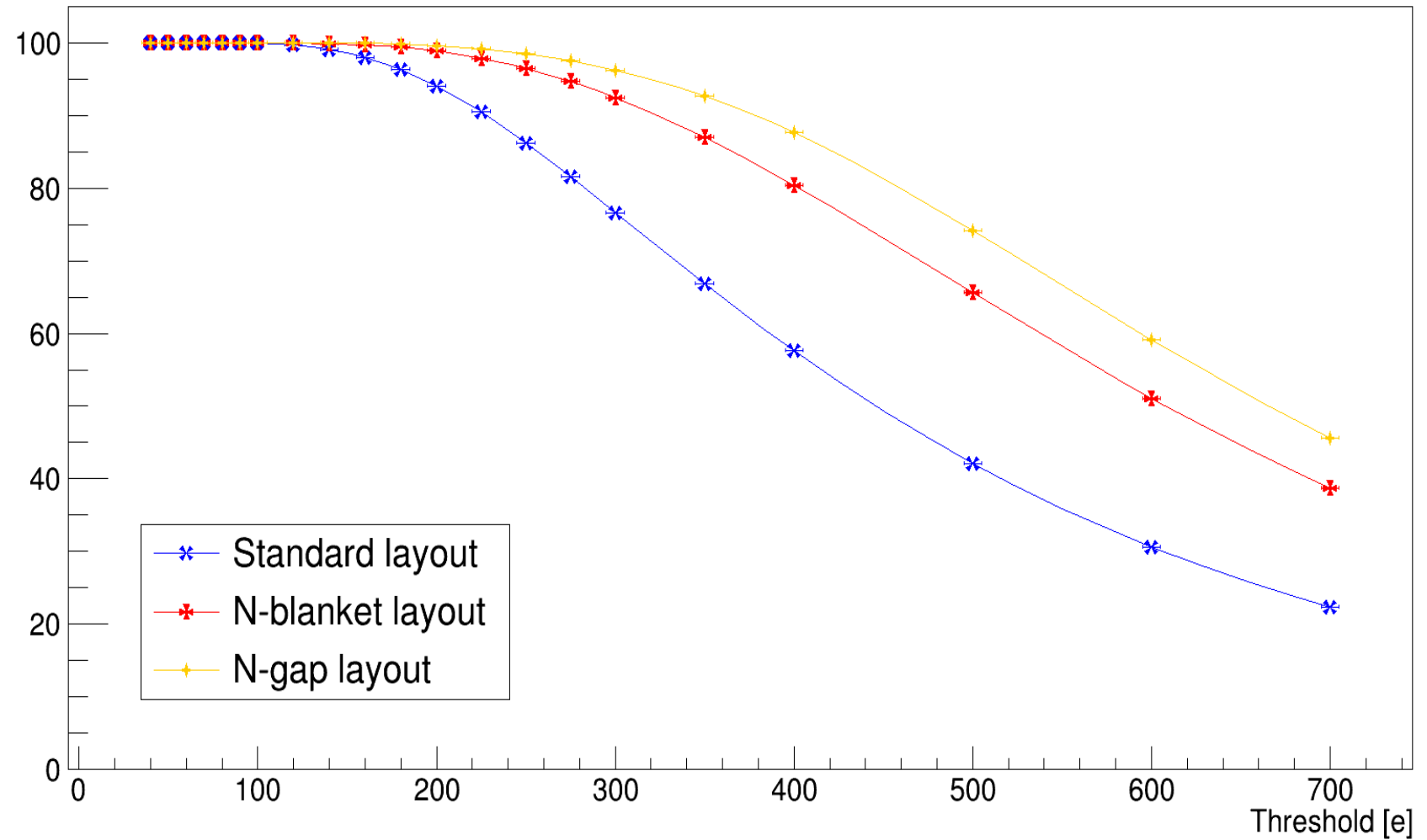
Fired = Number of Events that fired a pixel

Events = Number of particles shot

Number of events ~ 500000

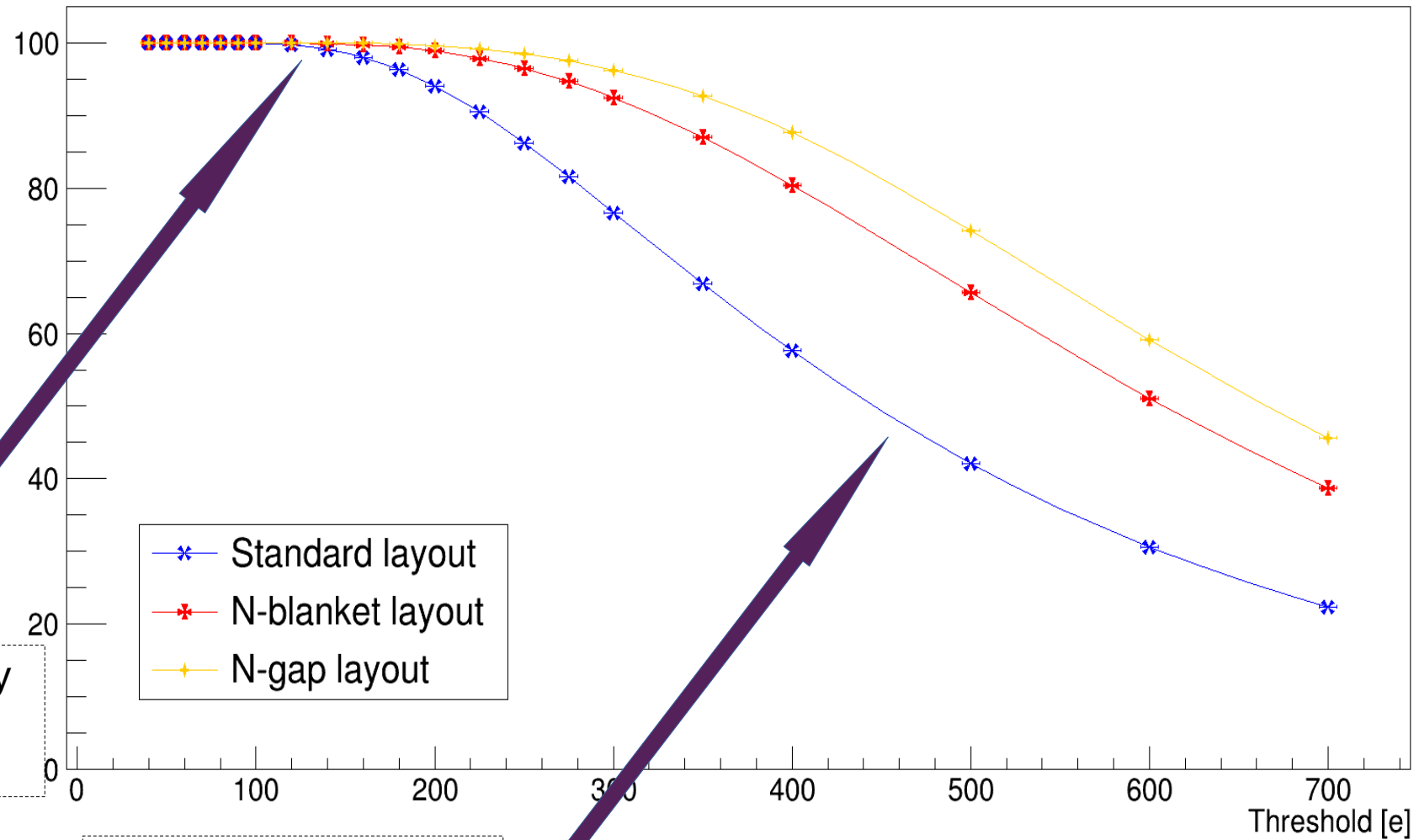
Layouts Comparison

Average Efficiency [%] 20x20 μm^2



Layouts Comparison

Average Efficiency [%] 20x20 μm^2



Maximum efficiency at low detection thresholds

Efficiency decreases as we increase the detection threshold

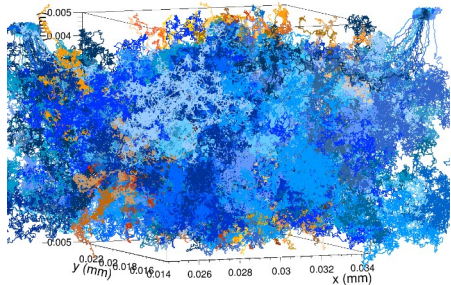
Layouts Comparison

Average Efficiency [%] 20x20 μm^2

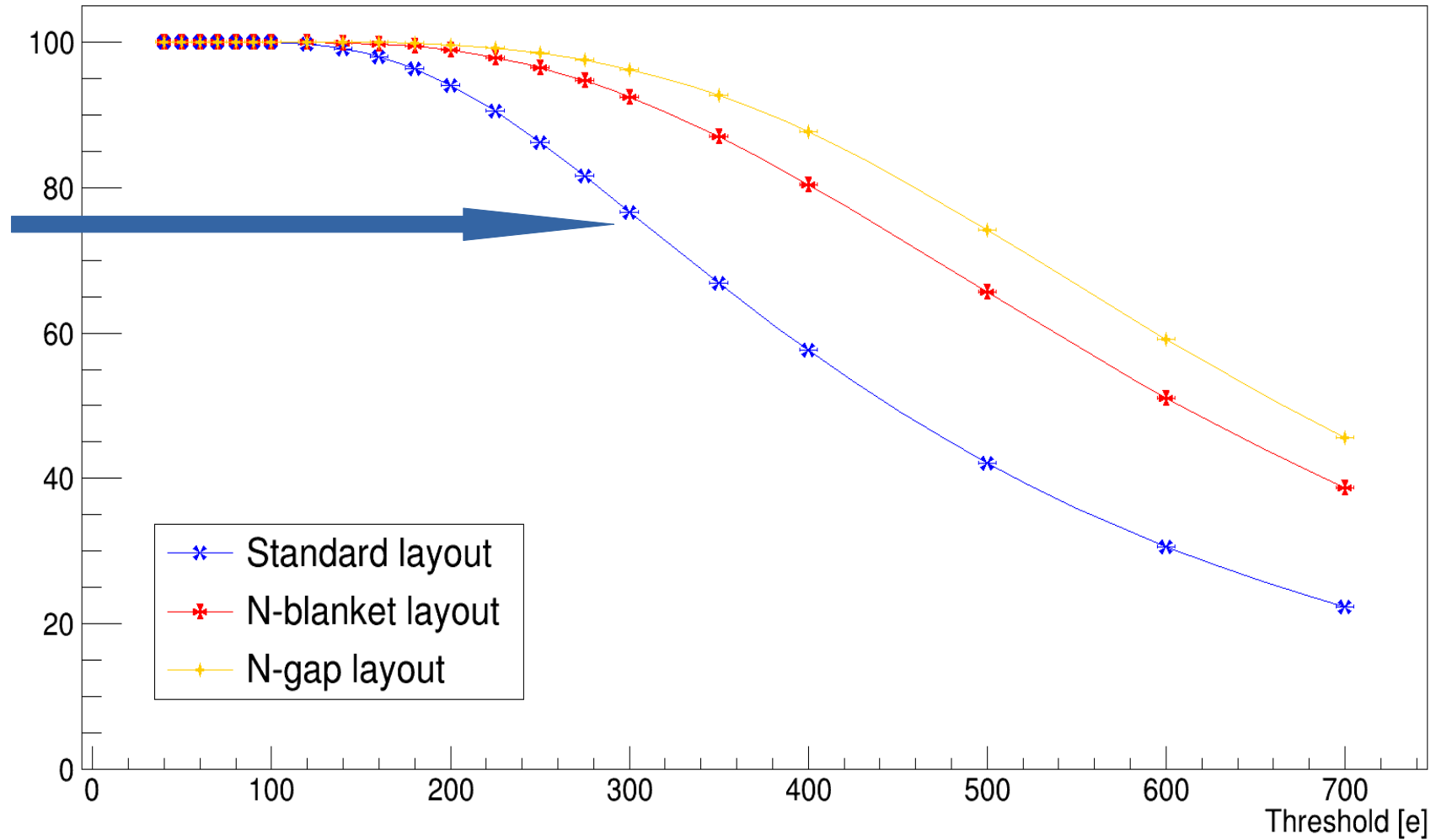
Efficiency decreases drastically for Standard layout



Diffusion

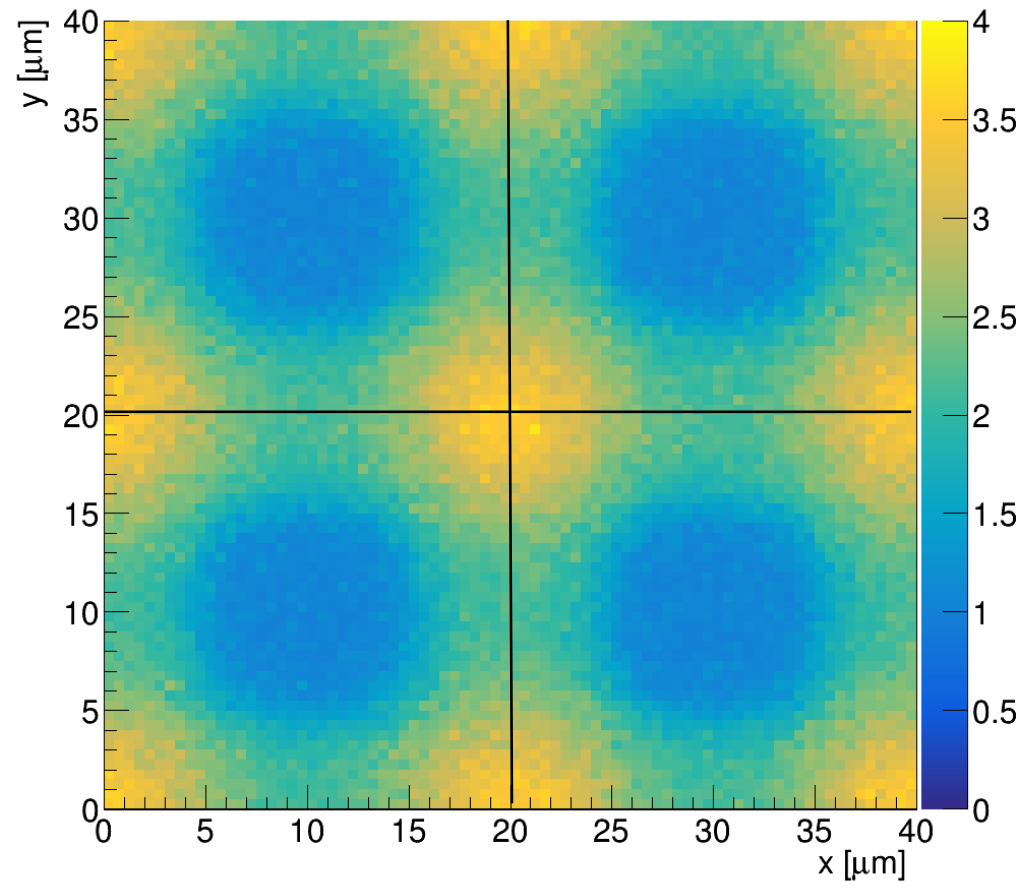


Standard layout linegraph



Cluster Size

4 Cluster Size Map , Threshold=80e



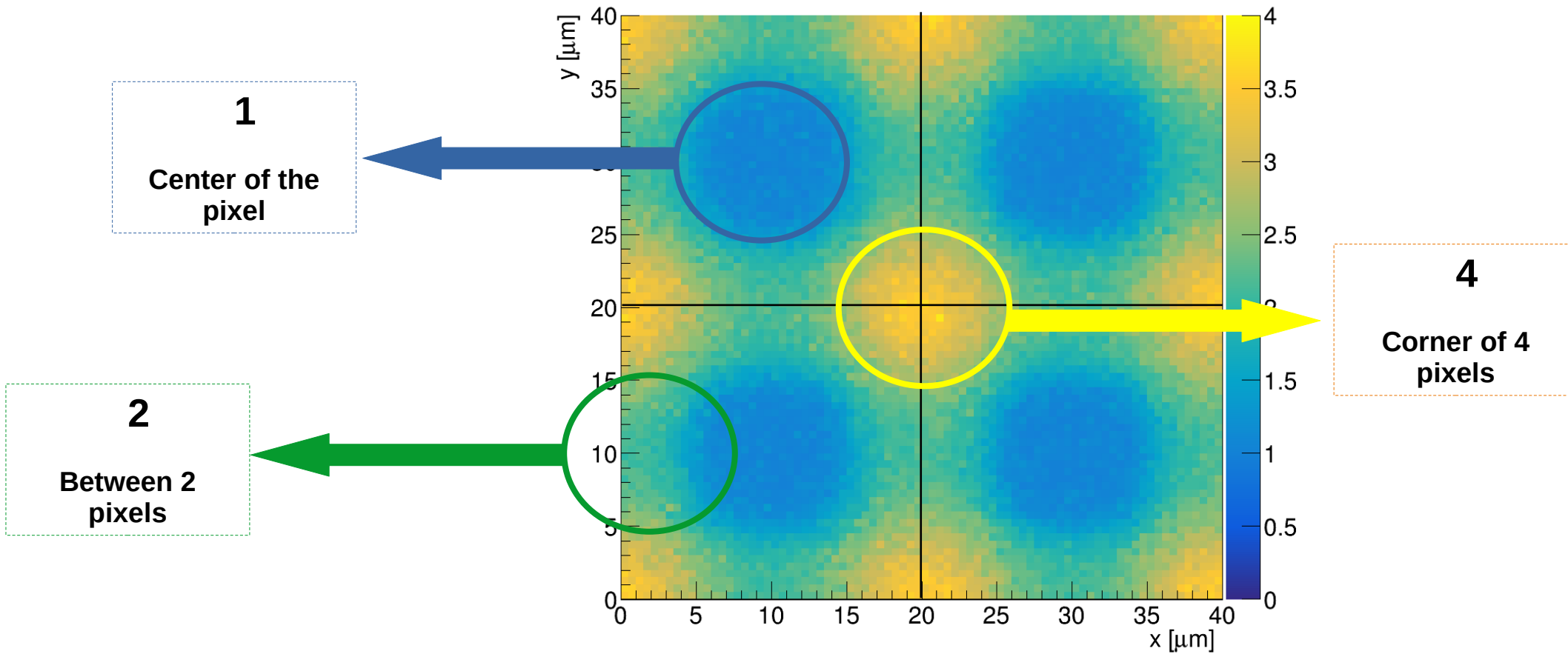
Number of Pixels fired per event.



A **high average cluster size**
(around 2) translates roughly to a
better (smaller) spatial
resolution.

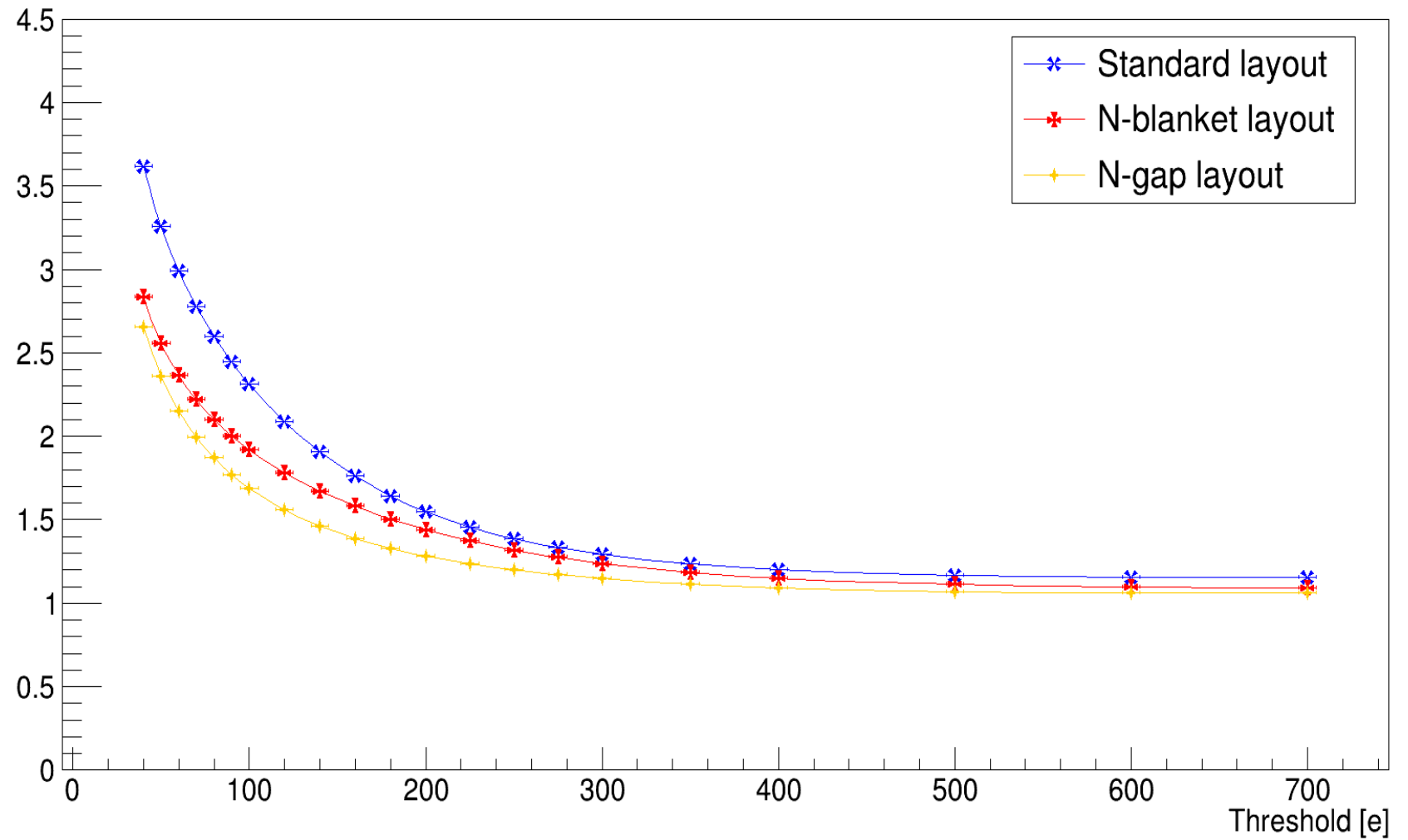
Cluster Size (Map of 4 adjacent pixels)

4 Cluster Size Map , Threshold=80e



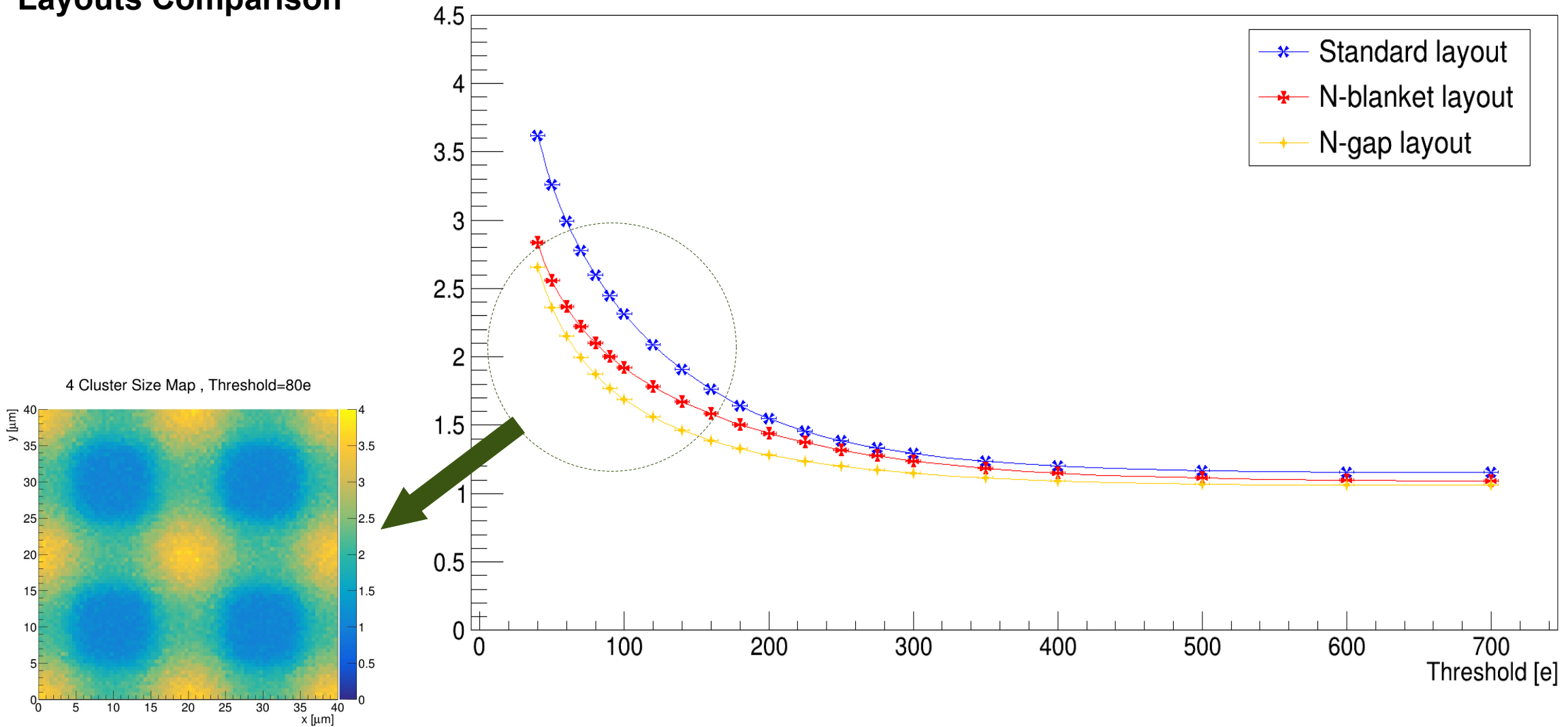
Layouts Comparison

Average Cluster Size $20 \times 20 \mu\text{m}^2$



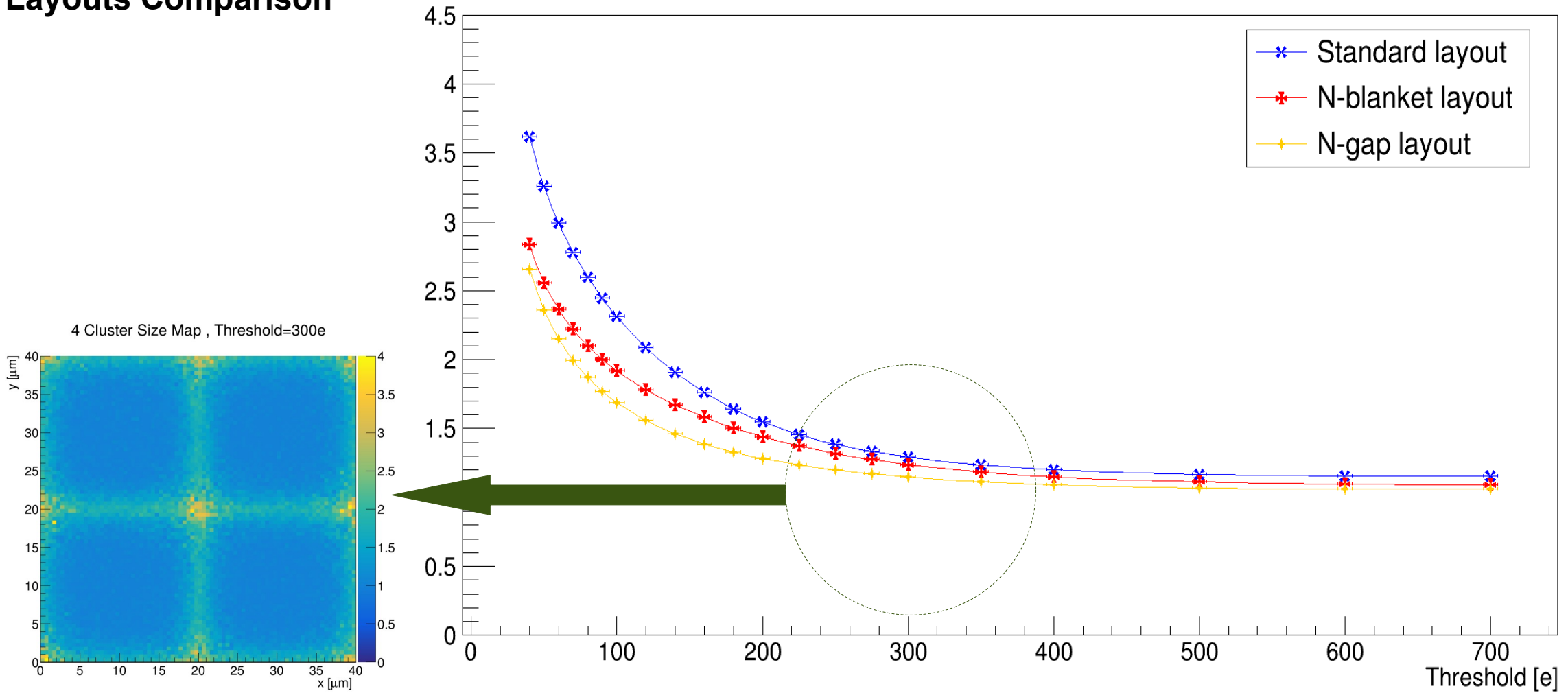
Layouts Comparison

Average Cluster Size 20x20 μm^2



Layouts Comparison

Average Cluster Size 20x20 μm^2



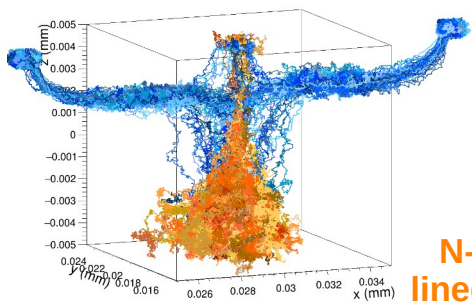
Average Cluster Size 20x20 μm^2

Layouts Comparison

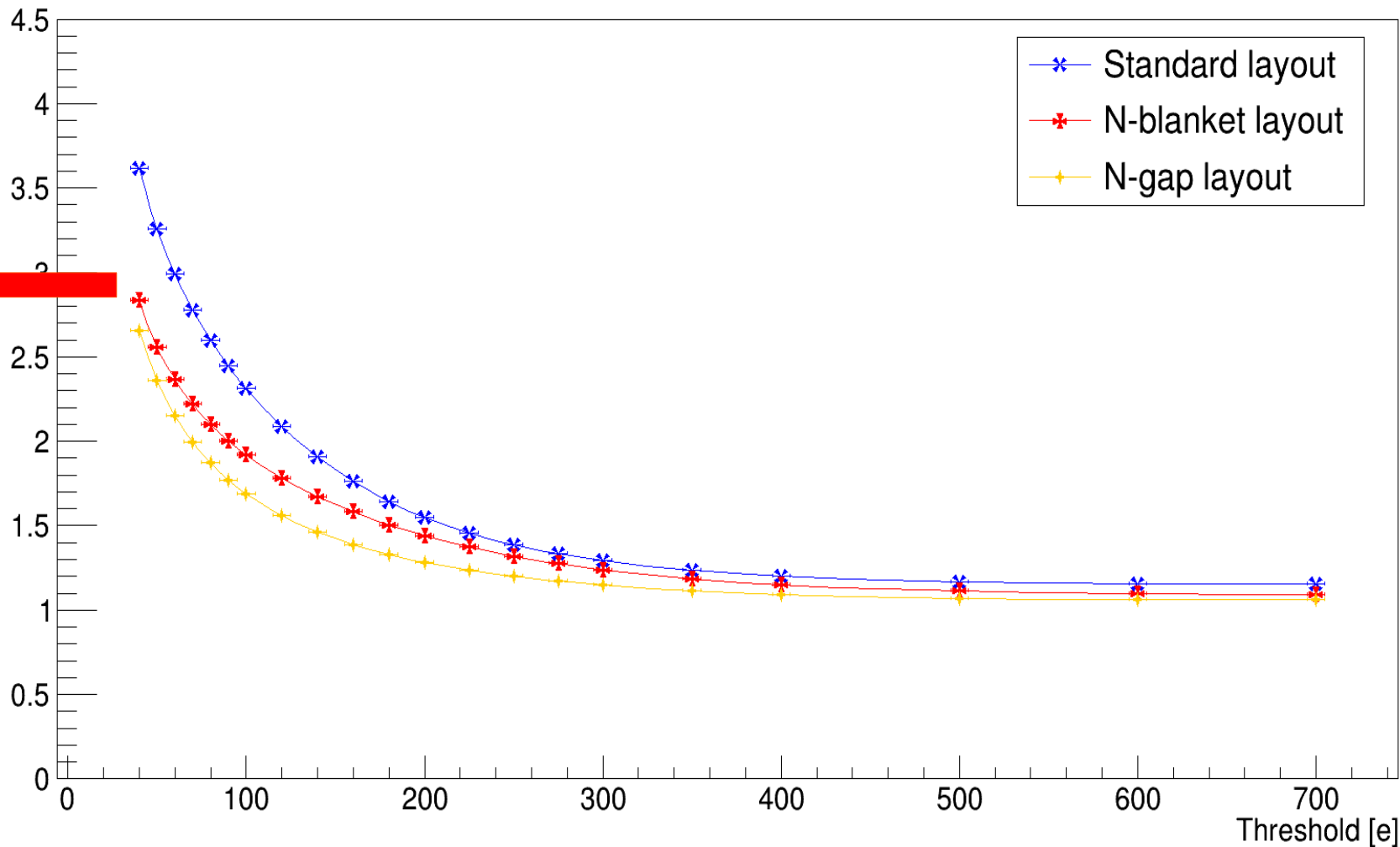
Smaller cluster size
N-blanket
&
N-gap



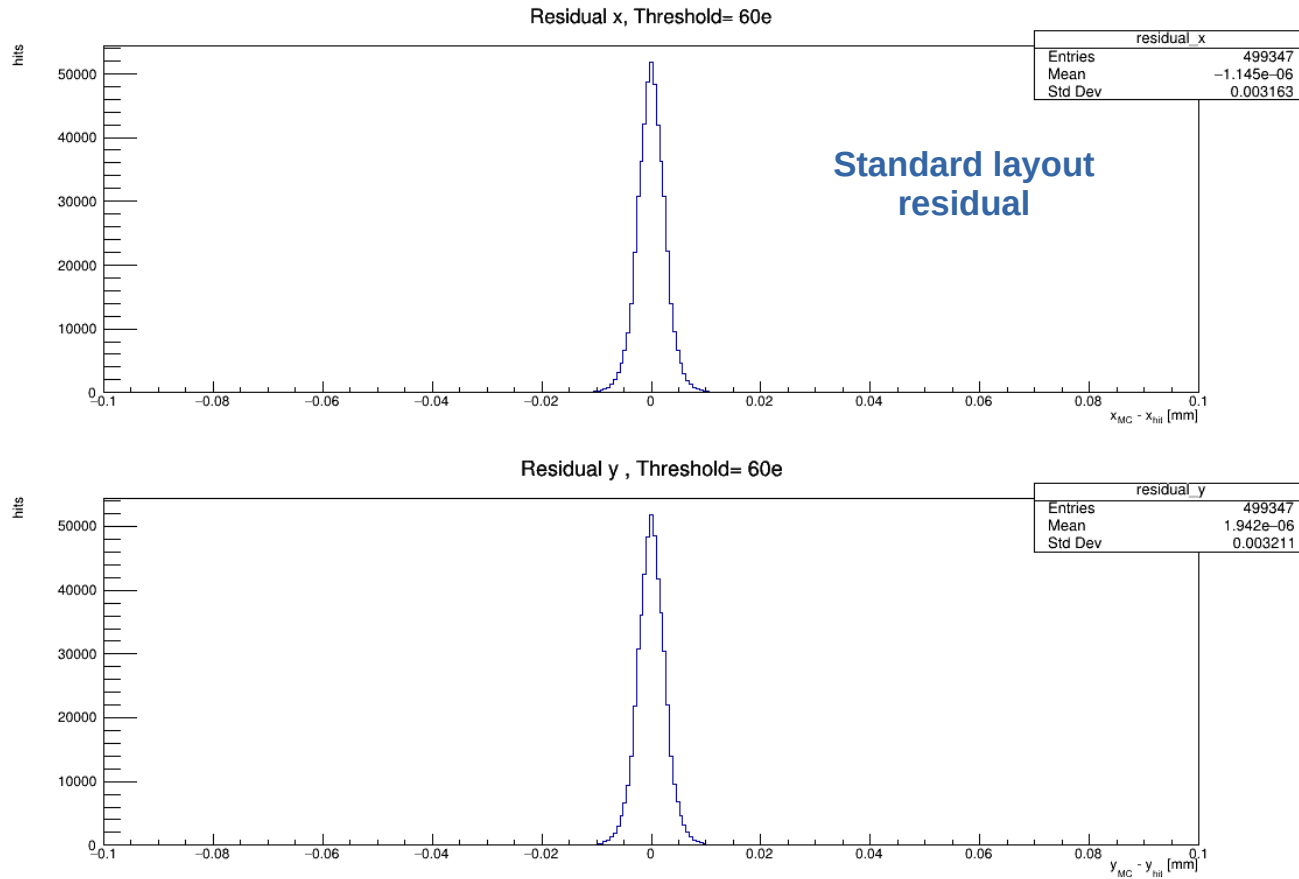
Drift



N-gap linegraph



Residuals ----> (Spatial Resolution)

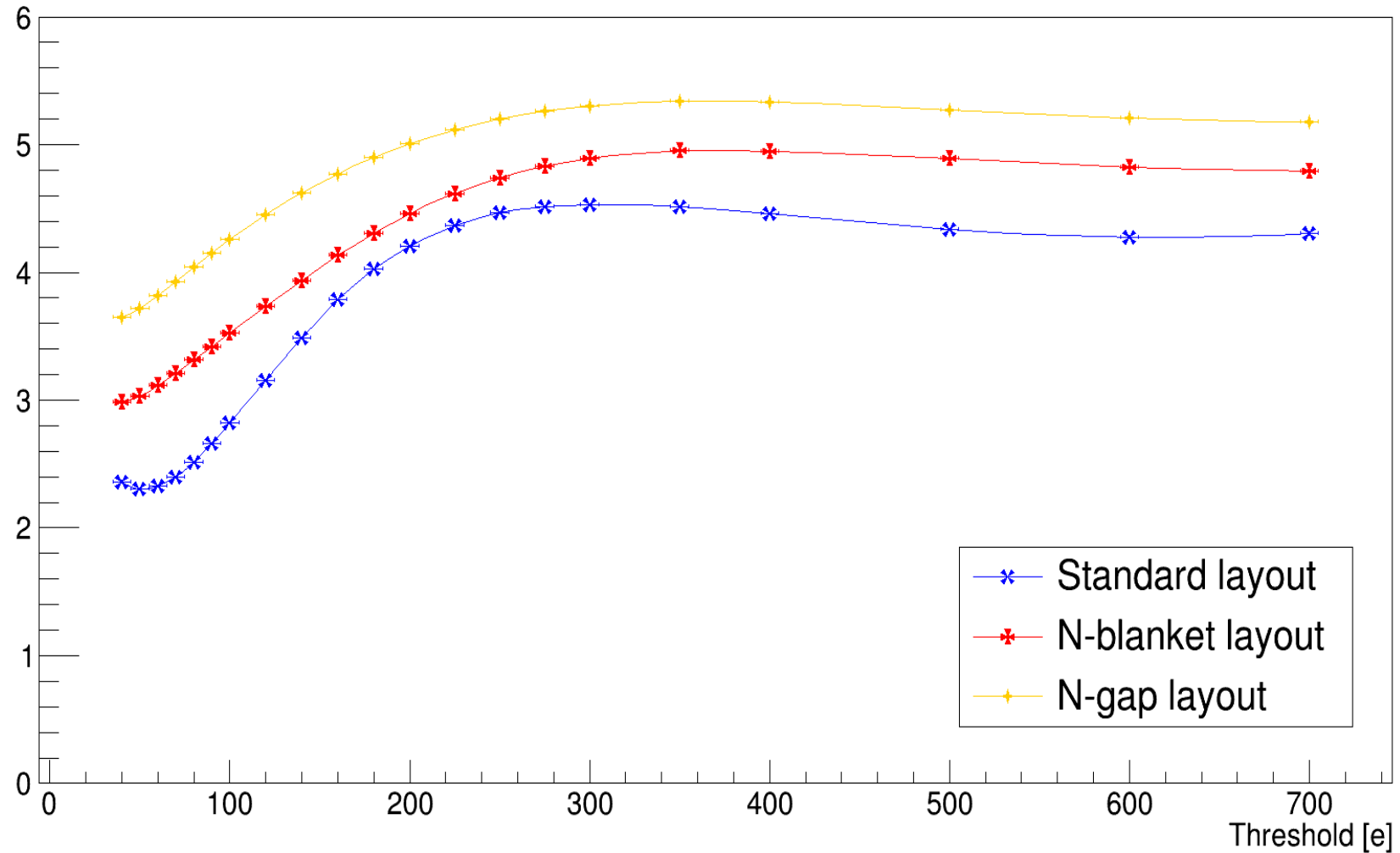


Residual: MC Particle incident position – Average cluster position



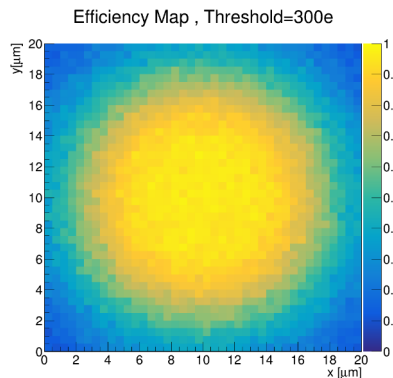
Spatial Resolution:
The RMS of the residual (or in this case ~ Standard Deviation)

Layouts Comparison

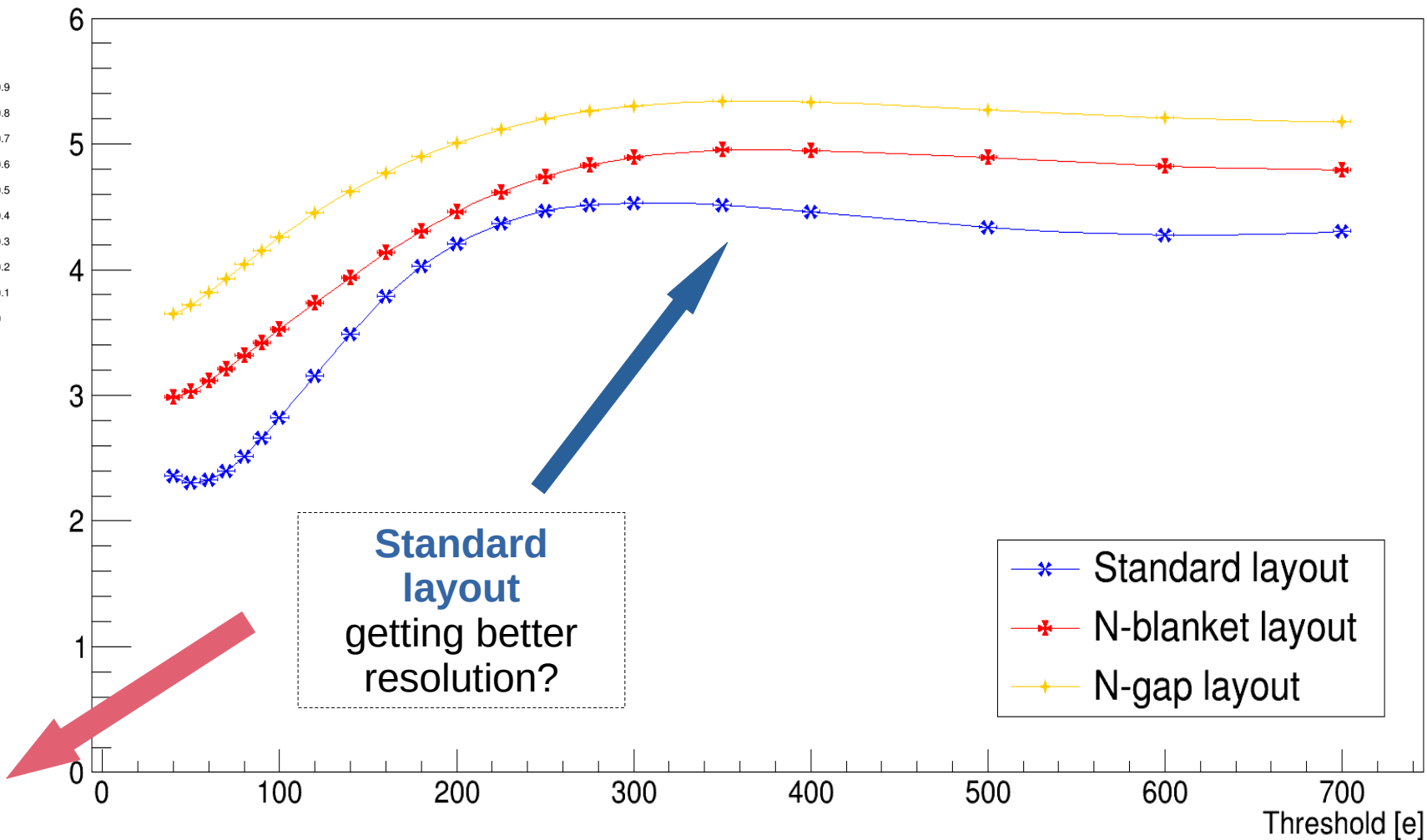
Average Spatial Resolution [μm] $20 \times 20 \mu\text{m}^2$ 

Layouts Comparison

Average Spatial Resolution [μm] $20 \times 20 \mu\text{m}^2$



Low efficiency!

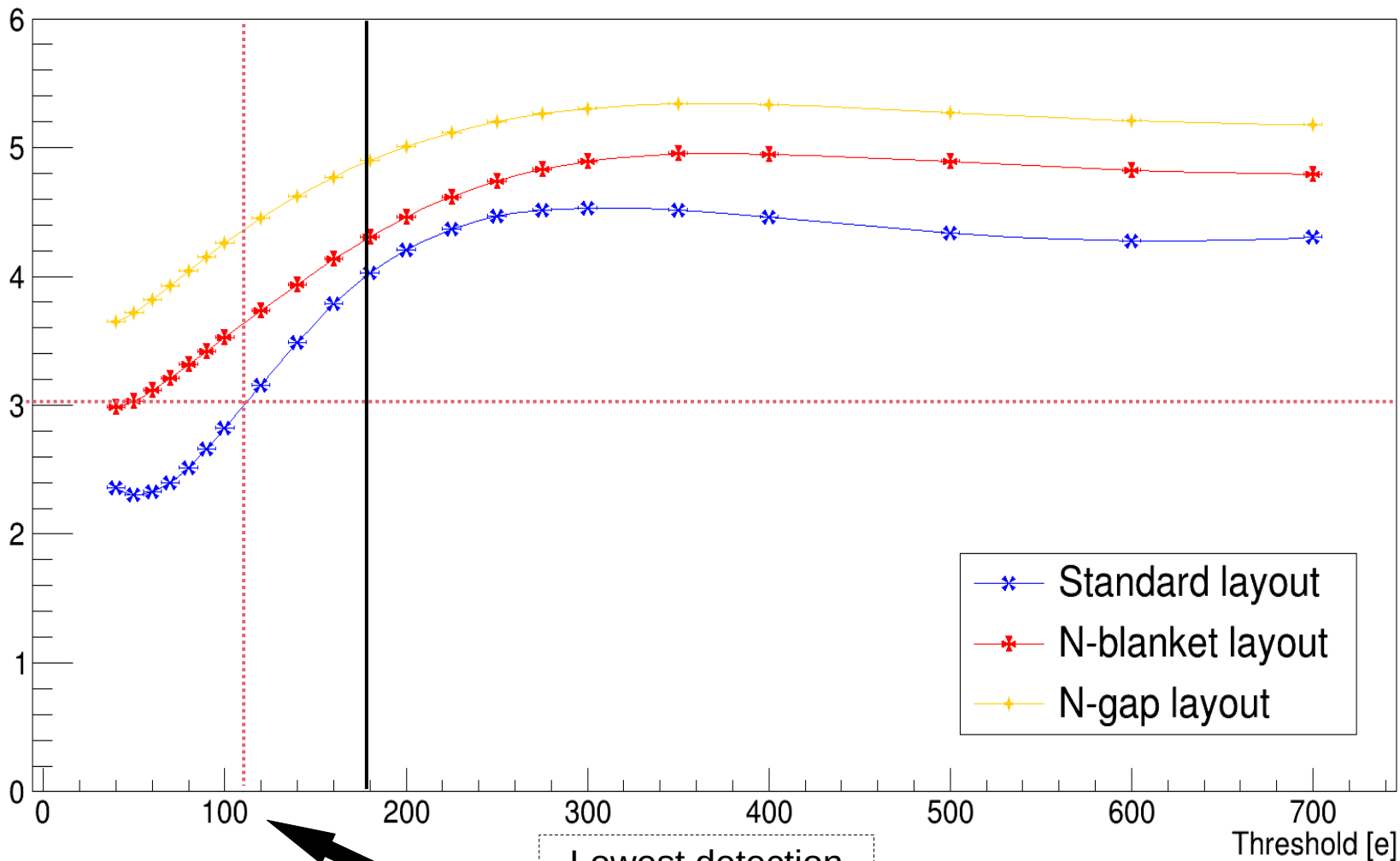


Standard layout getting better resolution?

- Standard layout
- N-blanket layout
- N-gap layout

Layouts Comparison

Average Spatial Resolution [μm] $20 \times 20 \mu\text{m}^2$



Desired resolution!
What else can we
change to achieve
this?

Lowest detection
threshold is not
always possible!
(Noise, electronics)

Summary

	Efficiency	Cluster Size (and Spatial Resolution)	Collection Time (Simulation Time)	Conclusion
Standard	<ul style="list-style-type: none"> • Harsh drop on efficiency 	<ul style="list-style-type: none"> • Largest of the three (thus smallest resolution) 	<ul style="list-style-type: none"> • ~ 25 ns (Due diffusion, longest simulation time) 	<ul style="list-style-type: none"> • Best resolution but efficiency might be concerning
N-blanket	<ul style="list-style-type: none"> • Good efficiency 	<ul style="list-style-type: none"> • Slightly larger than the N-gap (thus smaller than the N-gap) 	<ul style="list-style-type: none"> • > 25 ns (Drift dominated but for a minimum region) 	
N-gap	<ul style="list-style-type: none"> • Good efficiency 	<ul style="list-style-type: none"> • Smallest of them all (thus largest resolution) 	<ul style="list-style-type: none"> • > 25 ns (Drift dominated) 	<ul style="list-style-type: none"> • Great efficiency, worth looking into it.



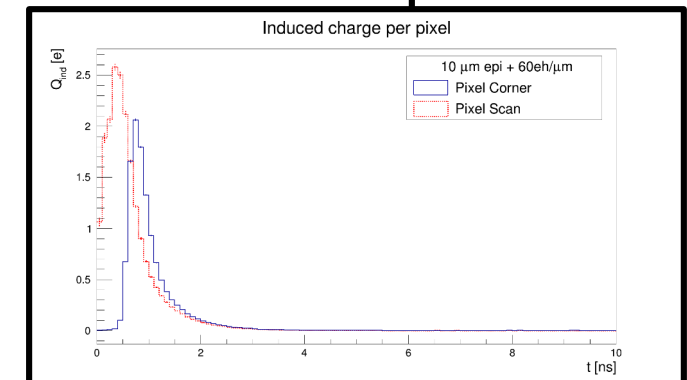
Conclusions and Next Steps!

Conclusions:

- Simulations provide many advantages and benefits.
- Improve from both parts on the simulation (both TCAD and MC).
- There still many parameters that can affect performance that could be worth looking into to achieve the desired capabilities.

Next Steps:

- Compare results with experimental data.
- Test the reproducibility, predictability and accuracy of the simulations.
- Fine tune and refine different parameters to achieve higher accuracy.
- Produce **Transient simulations** studies to understand the sensor response.



Thank you for your time!

Contact:

Deutsches Elektronen-Synchrotron

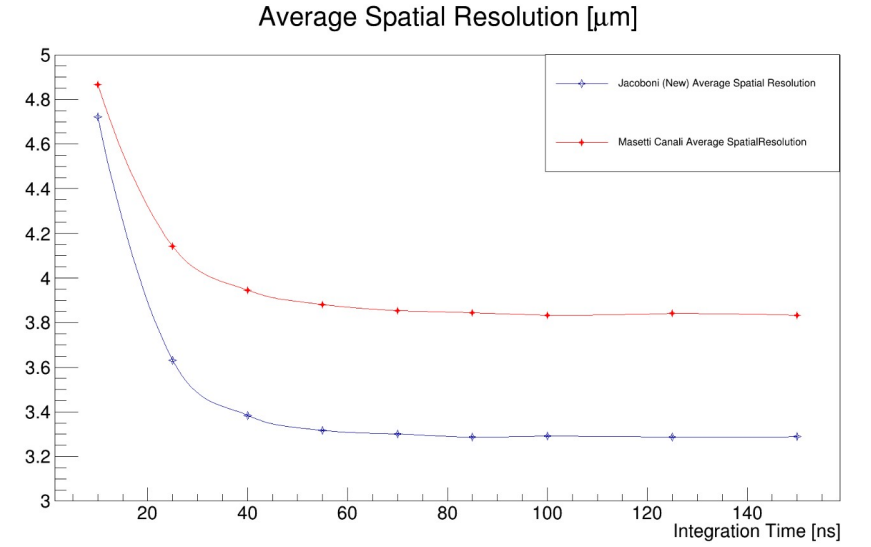
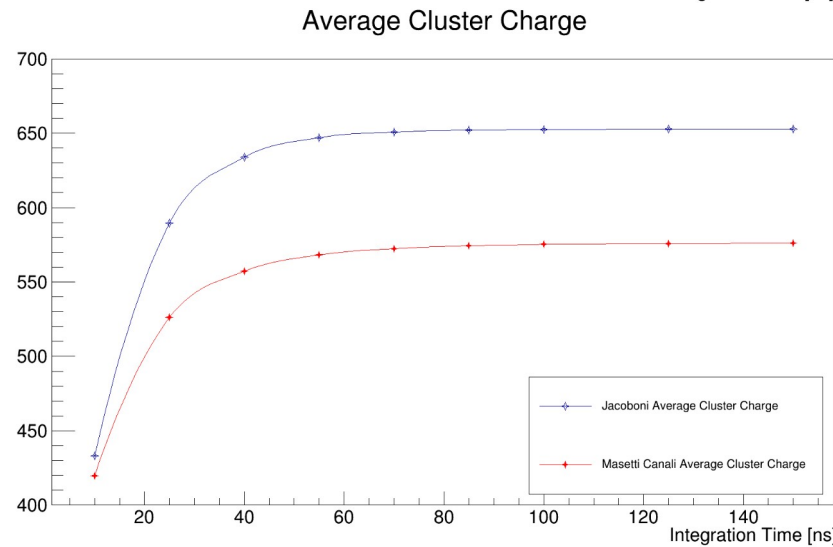
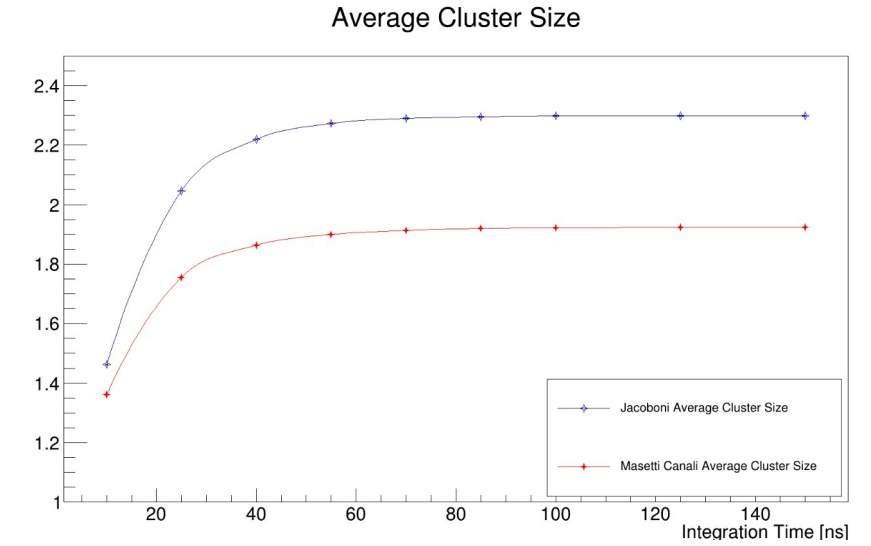
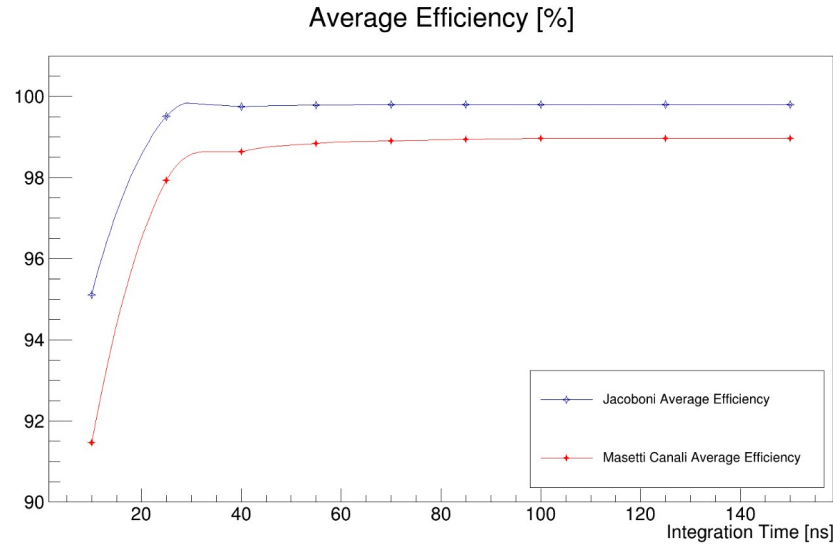
ATLAS

manuel.del.rio.viera@desy.de

Backup

Integration time (Total Simulation time)

- Non-doping dependent Mobility Model
- Doping dependent Mobility Model



Simulation Results (Comparison between Mobility Models)

A **mobility model** refers to a model describing the electric field and doping concentration dependence of the charge carrier velocity.

We will compare the following two mobility models:

Jacoboni-Canali

$$\mu(E) = \frac{v_m}{E_c} \frac{1}{(1 + (E/E_c)^\beta)^{1/\beta}},$$

Masetti-Canali

$$\mu(E, N) = \frac{\mu_m(N)}{(1 + (\mu_m(N) \cdot E/v_m)^\beta)^{1/\beta}}$$

Note that Jacoboni model does not depend on the **doping concentration** (explicitly) while Masetti does

Linegraphs

