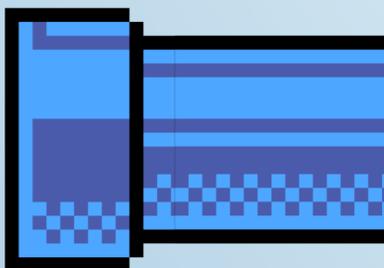
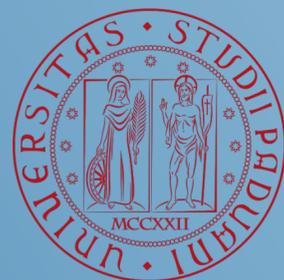


# Towards the optimization of a Muon Collider Calorimeter

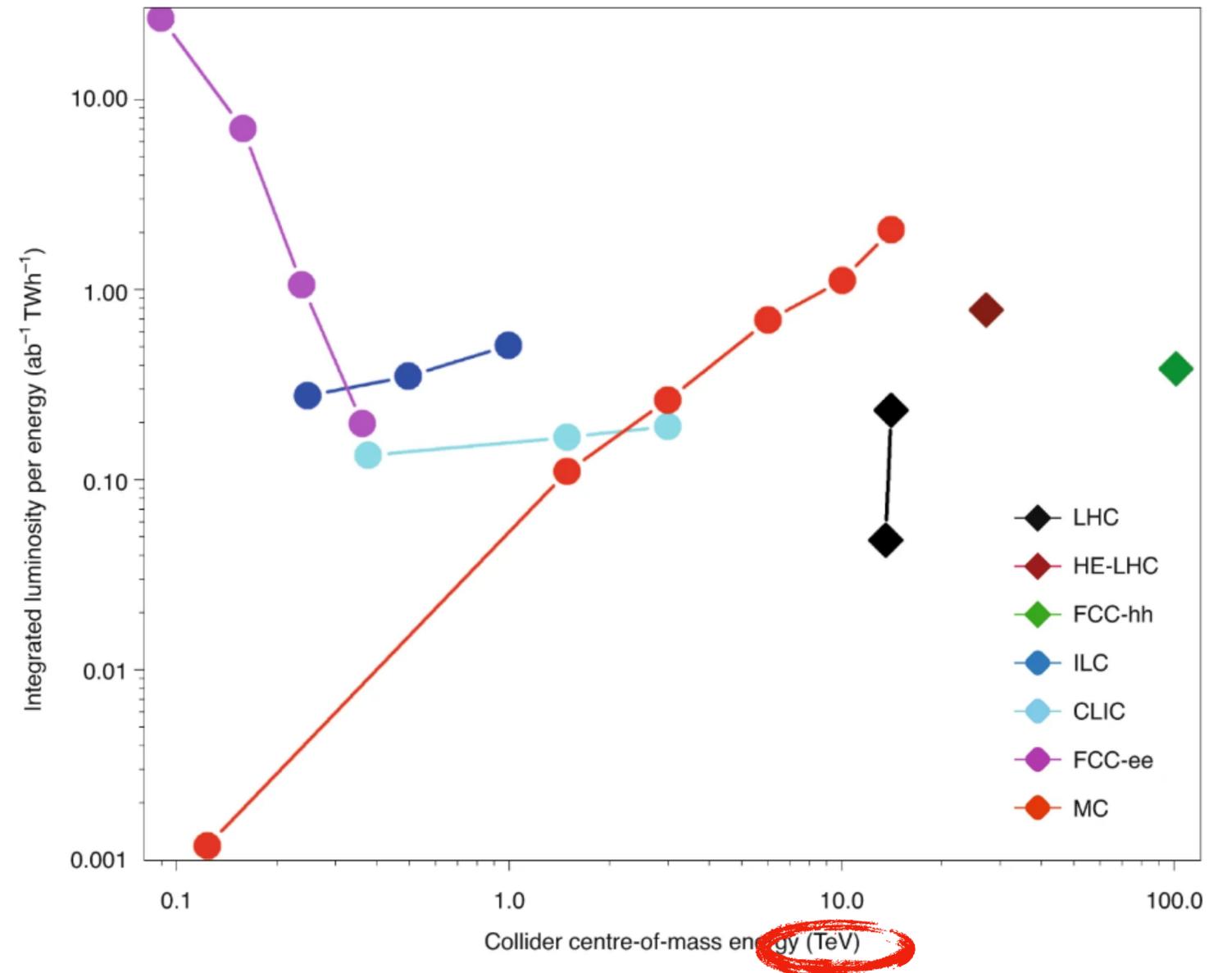
F. Nardi, J. Donini, T. Dorigo, J. Kieseler, A. De Vita, Abhishek, M. Aehle, M. Awais, A. Breccia, R. Carroccio, L. Chen, N. R. Gauger, R. Keidel, E. Lupi, X.T. Nguyen, F. Sandin, K. Schmidt, P. Vischia, J. Willmore



# Introduction

## Why a Muon Collider?

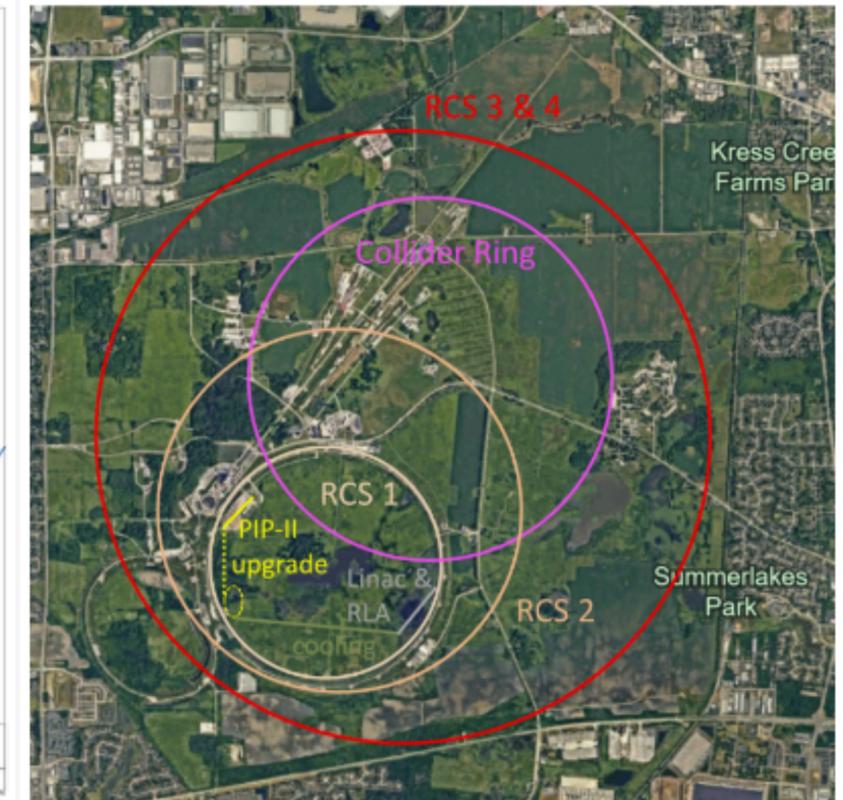
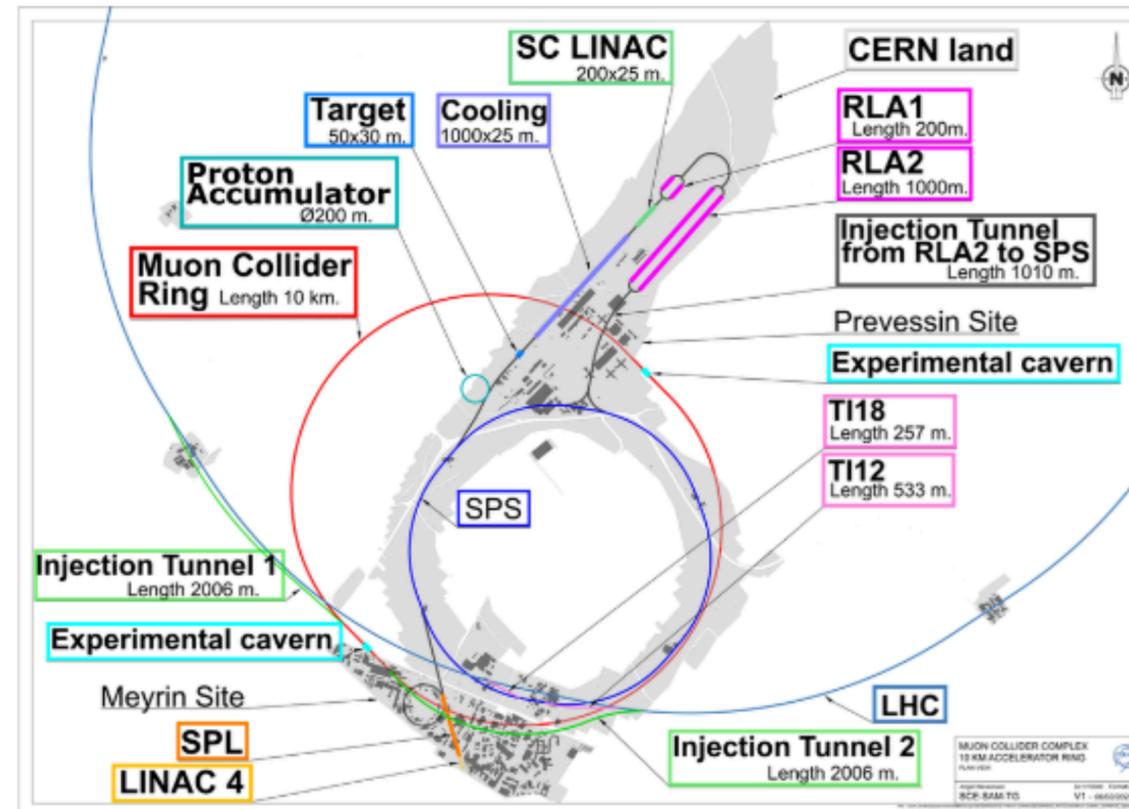
- Luminosity increases with center-mass energy
  - Competitive with LINACs
  - Most 'physics-per-dollar' potential
- Heavier than electrons: less radiative losses -> Higher center mass energies
- Lepton Collider: no pile-up effects
- Rather old concept, regained interest within the post-LHC debate (Snowmass, European Strategy): possibility to have a lepton collider working at TeV scale
- Higgs Factory
  - $\sigma(\mu\mu \rightarrow H) \approx 40000 \sigma(ee \rightarrow H)$
- Possibility for BSM measurements



# Introduction

## Why a Muon Collider?

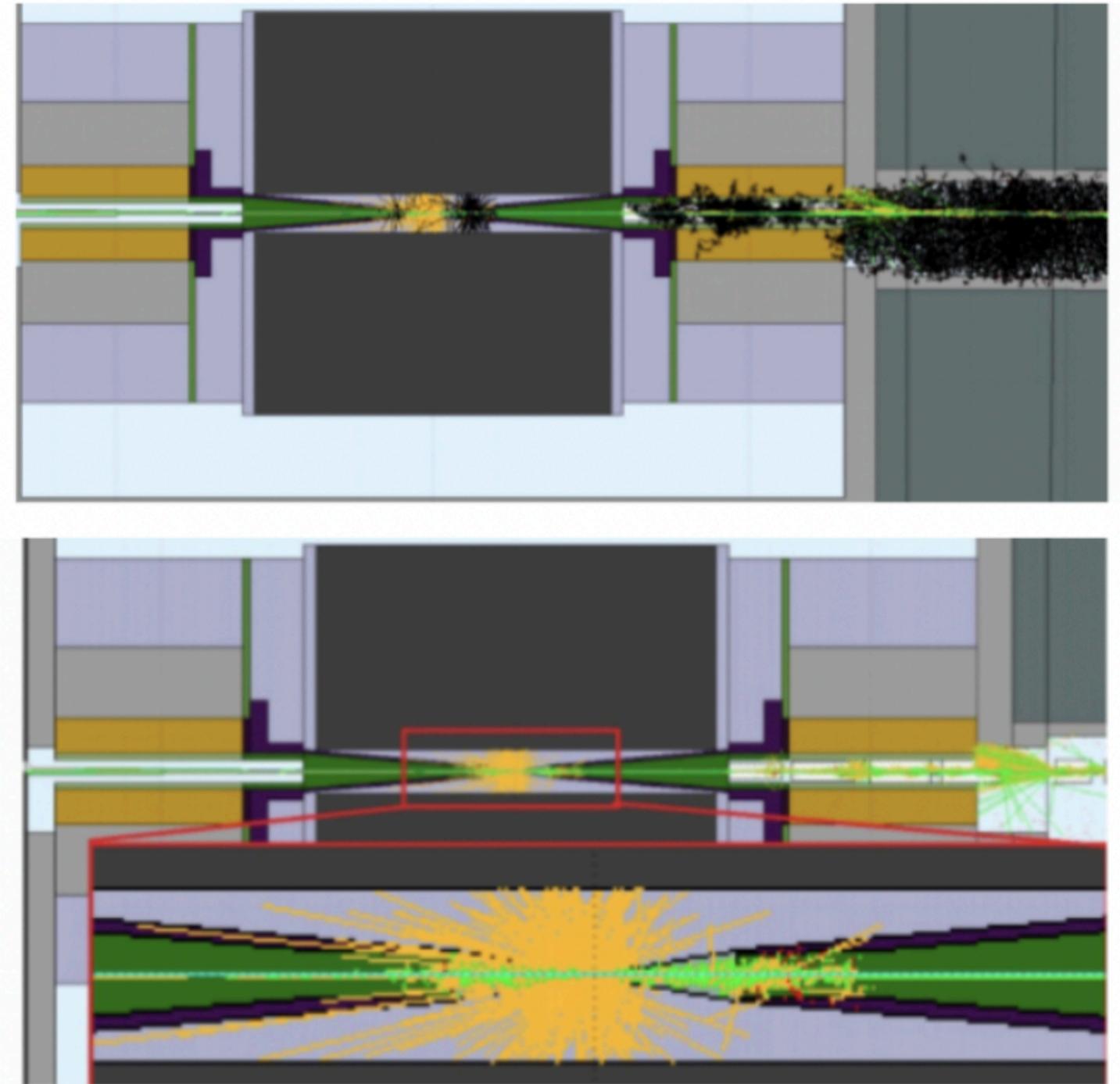
- Beamline plans for CERN(left) and Fermilab(right).
- O(10km) rings for TeV-scale collision



# What

## Summary of the problem

- Finite lifetime of the muon ( $2.2\mu\text{s}$ ) implies a cloud of high-energy decay product along the beamline, which interferes with the instrumentation (Beam-Induced Background - BIB)
- During preliminary Machine-Detector Interface design, a double-cone nozzle has been included to shield the detector from BIB radiation

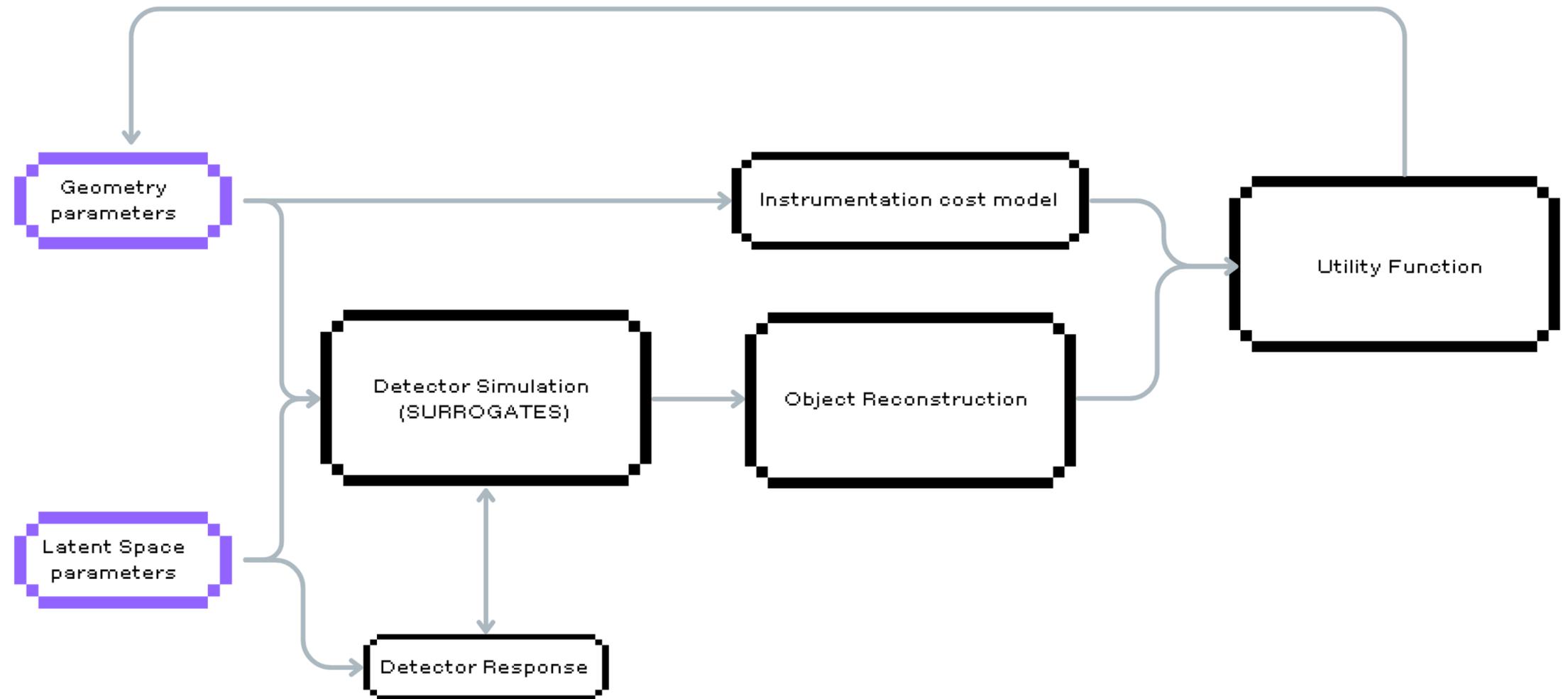


Visualizations from FLUKA BIB simulation. Black: neutrons, other: photons

# What

## Pipeline scheme

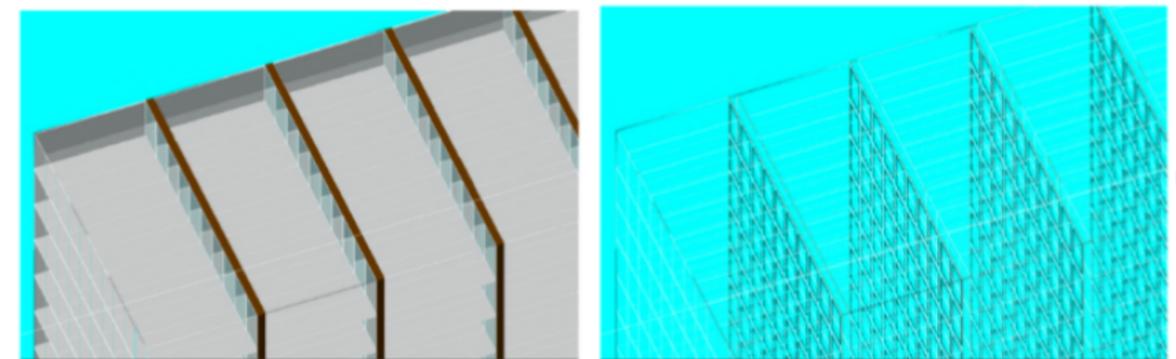
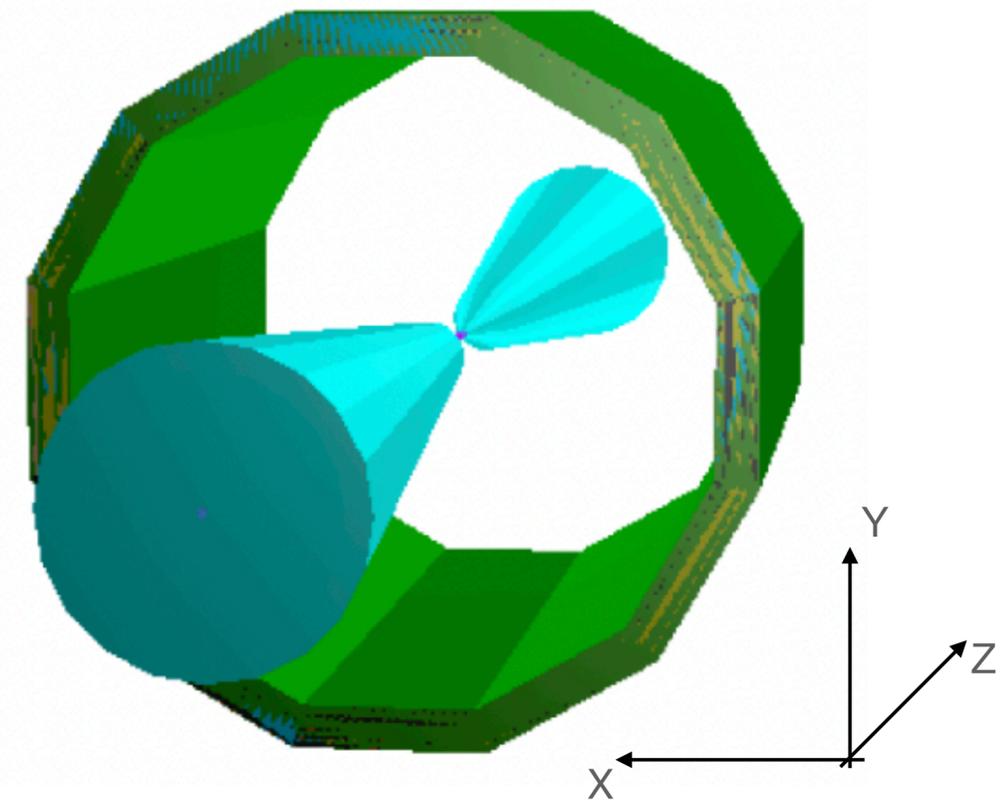
- End objective: design optimization study approached with AD techniques
- Development of a pipeline to propose an optimal configuration in terms of **signal-to-background discrimination** and **instrumentation cost**
- Use single monochromatic photons as benchmark



# What

## CRILIN: Reference design

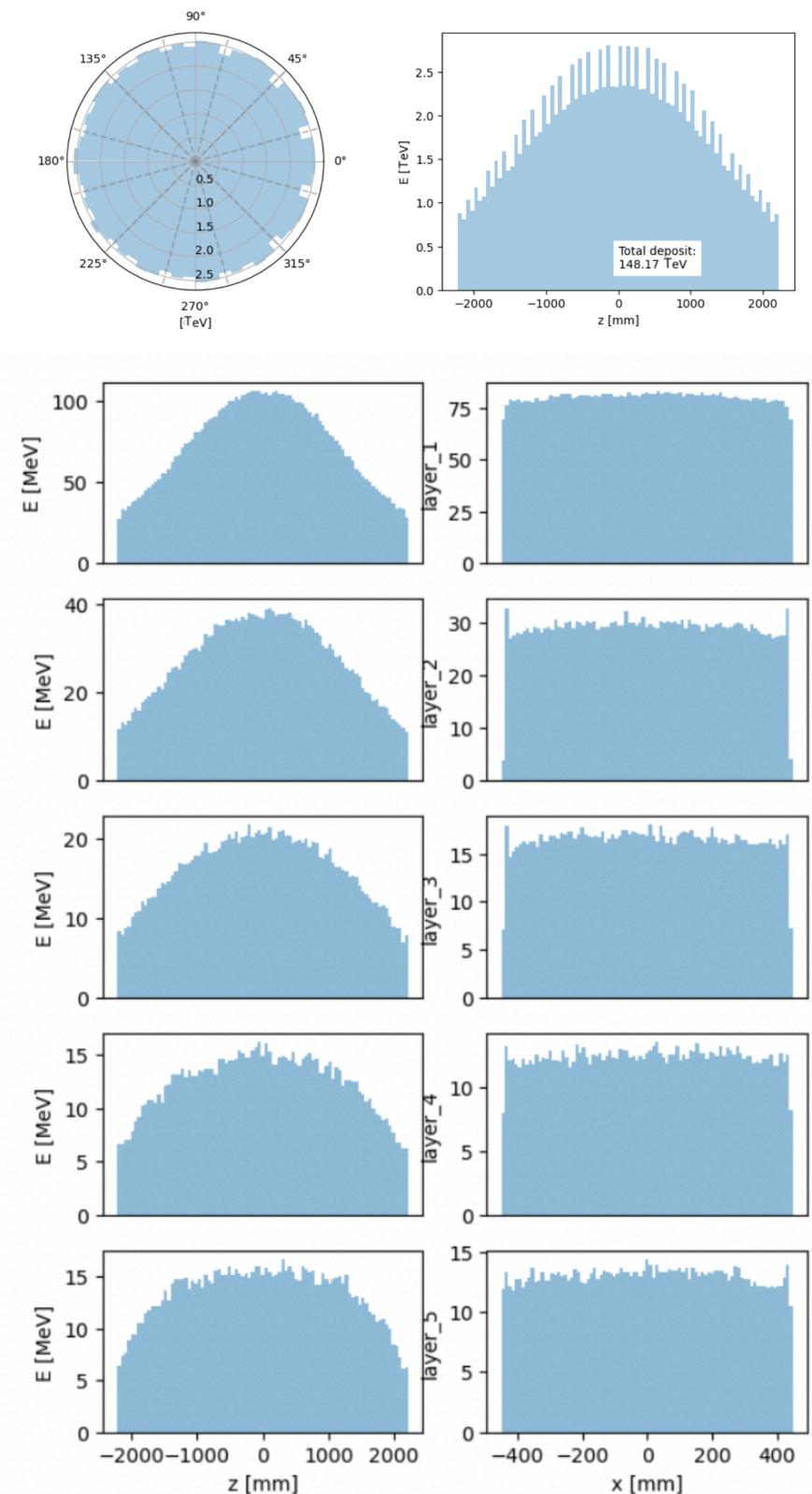
- Reference design chosen for our studies is CRILIN for the Electromagnetic Calorimeter (ECal)
- Array of  $1 \times 1 \times 4.5 \text{ cm}^3$   $\text{PbF}_2$  voxels, arranged in a dodecagonal prism
- 5 layers per wedge
- Modular design, easy to modify and rearrange



# Modules

## BIB Generation

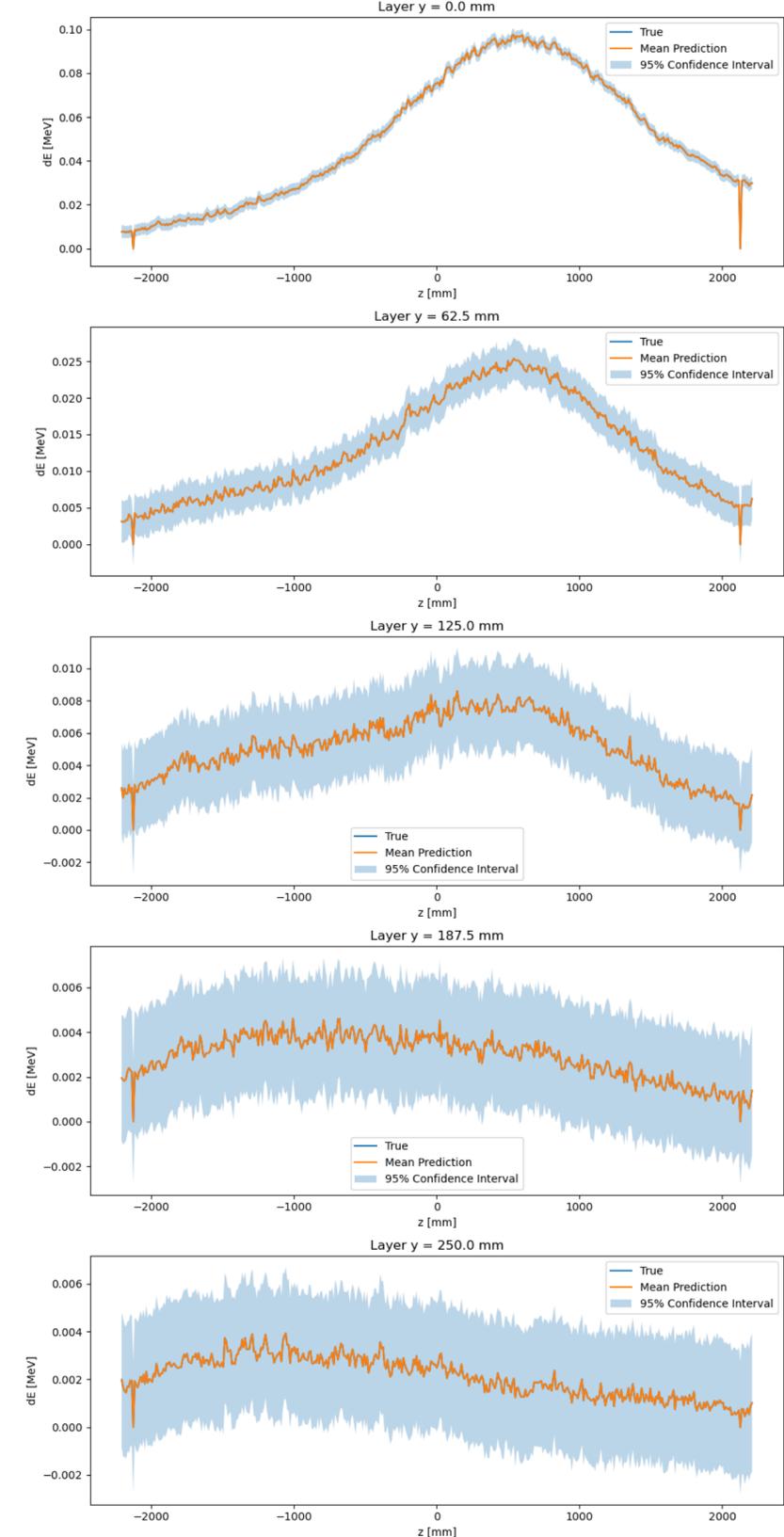
- Starting from a 750GeV FLUKA beam simulation, fed to a Geant4 detector model.
- Muon decays within 5m from interaction point
- Symmetrized to simulate contributions from both sides
- Cylindrical symmetry allows to limit the problem to a single wedge



# Modules

## BIB Generation

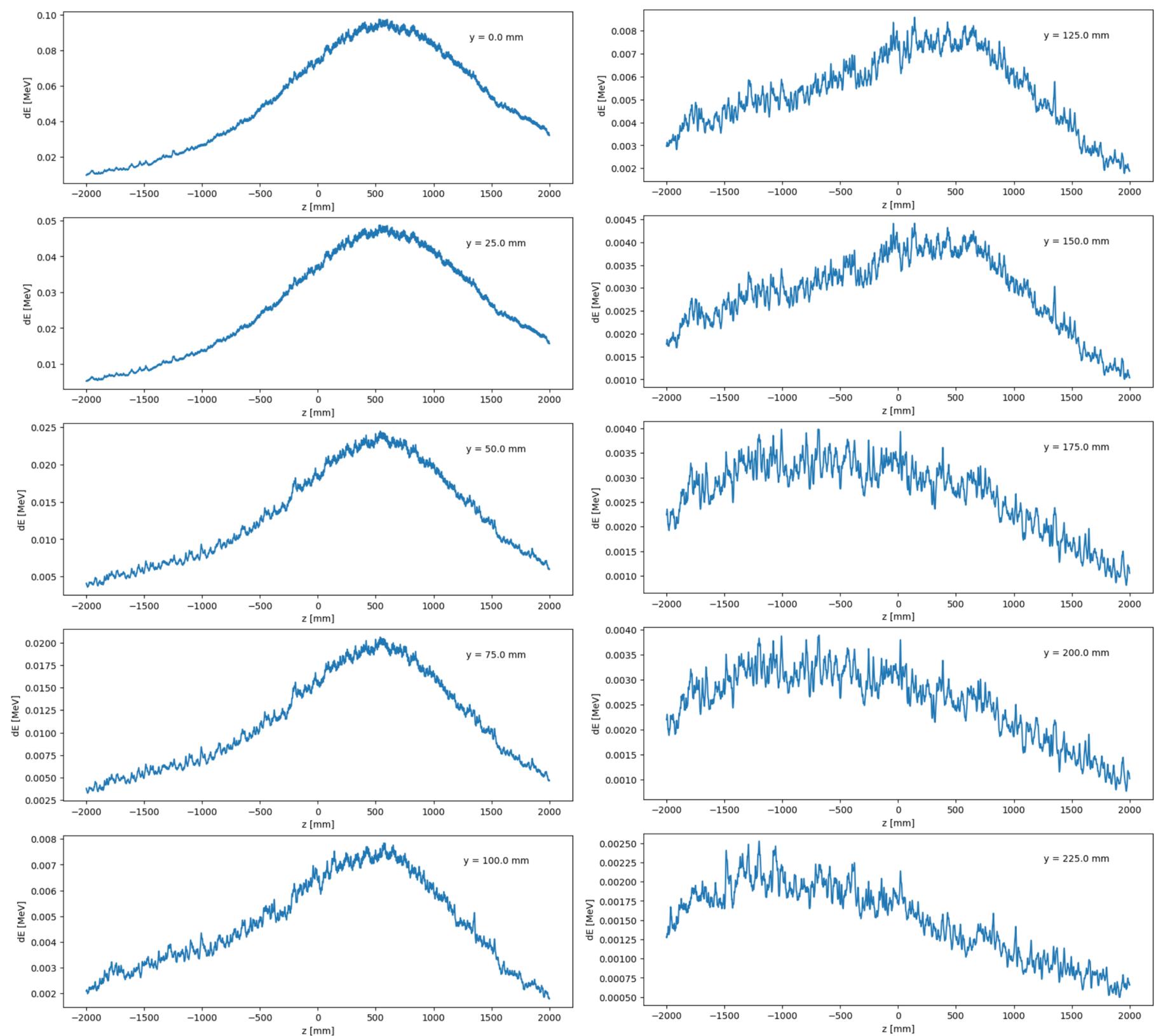
- Model the BIB flux across the 5 layers by training a Gaussian Process
- Predicts BIB energy density given a  $(z,y)$  coordinate pair - uniform along the  $x$  direction
- Main motivation for a GP is ability to interpolate between layers



# Modules

## BIB Generation

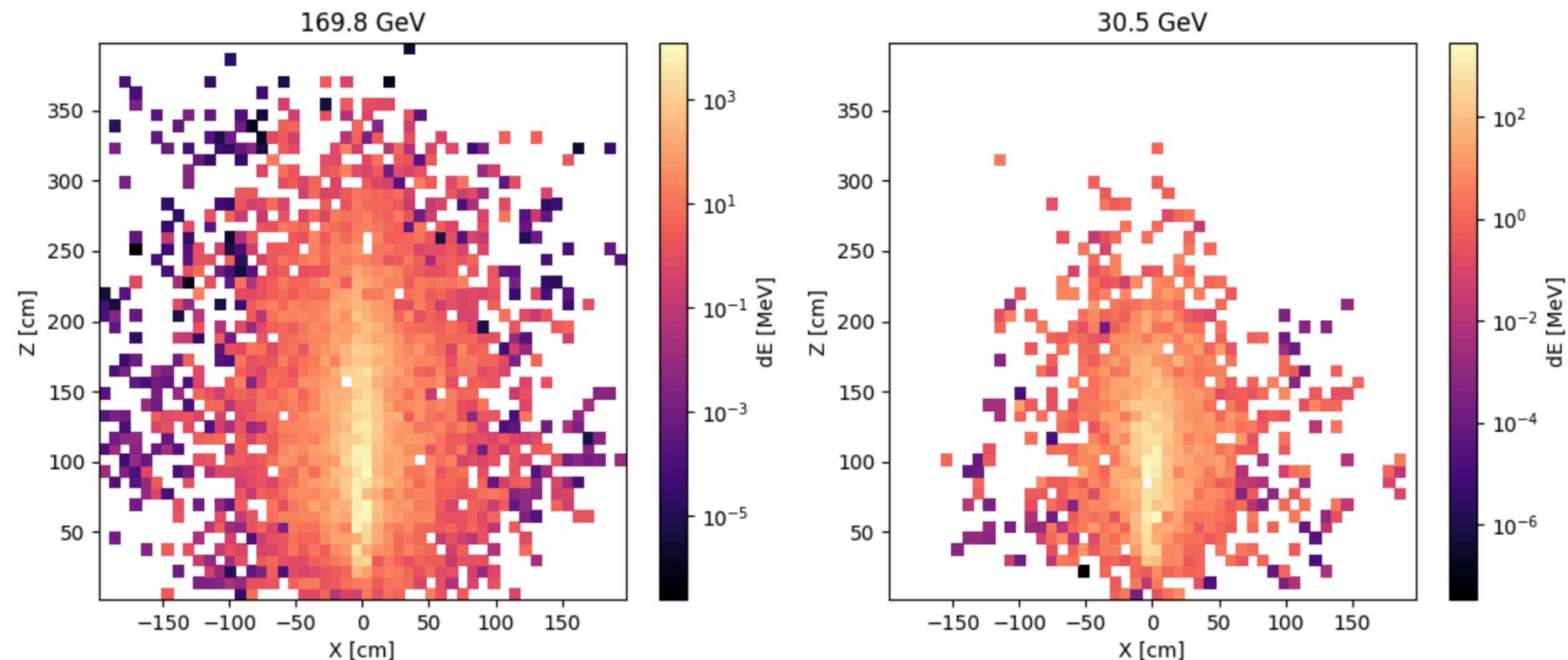
- Simulation is heavy ( $O(\text{TB})$  for single event), and does not provide information on intra-layer BIB depositions
- Validation: total deposit on the detection area matches simulation
- Shapes and magnitudes are reasonable, allows us to work



# Modules

## Shower Generation

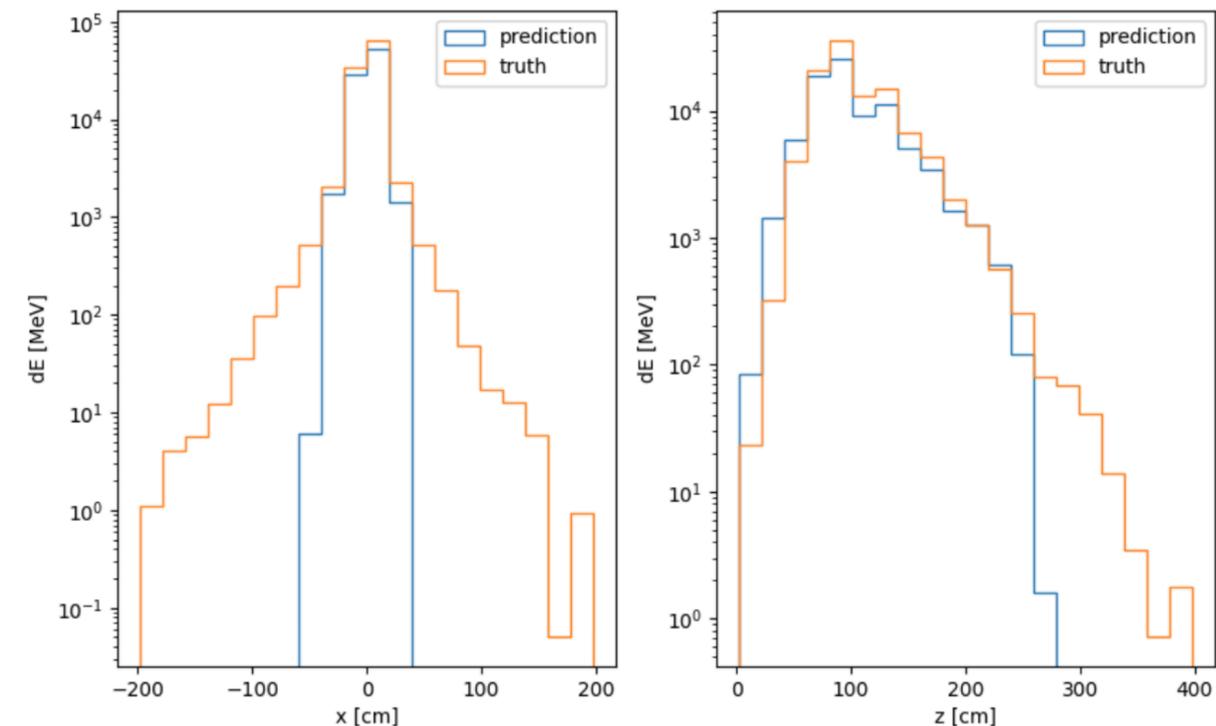
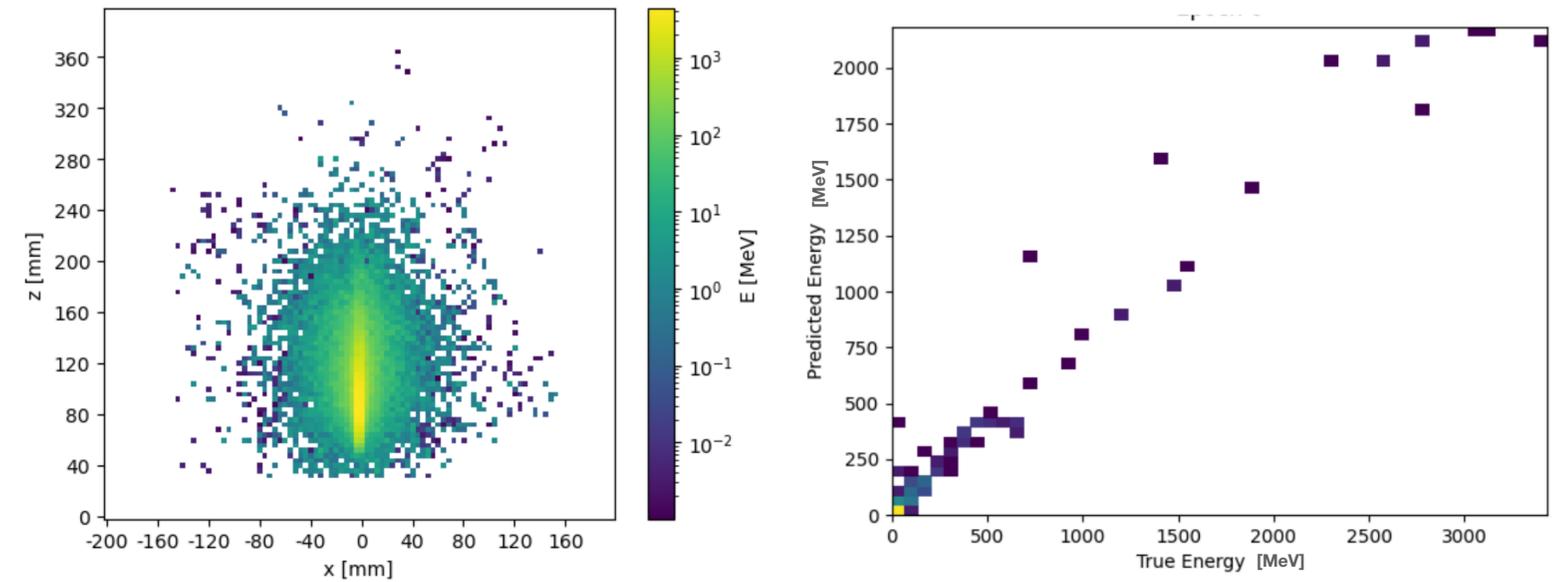
- For signal events we chose single monochromatic photons, energy uniformly distributed in  $[10, 175]$  GeV.
- Use Geant4 to produce a dataset of 10k photons entering orthogonally in a PbF2 block.
- Saved as 2D images



# Modules

## Shower Generation

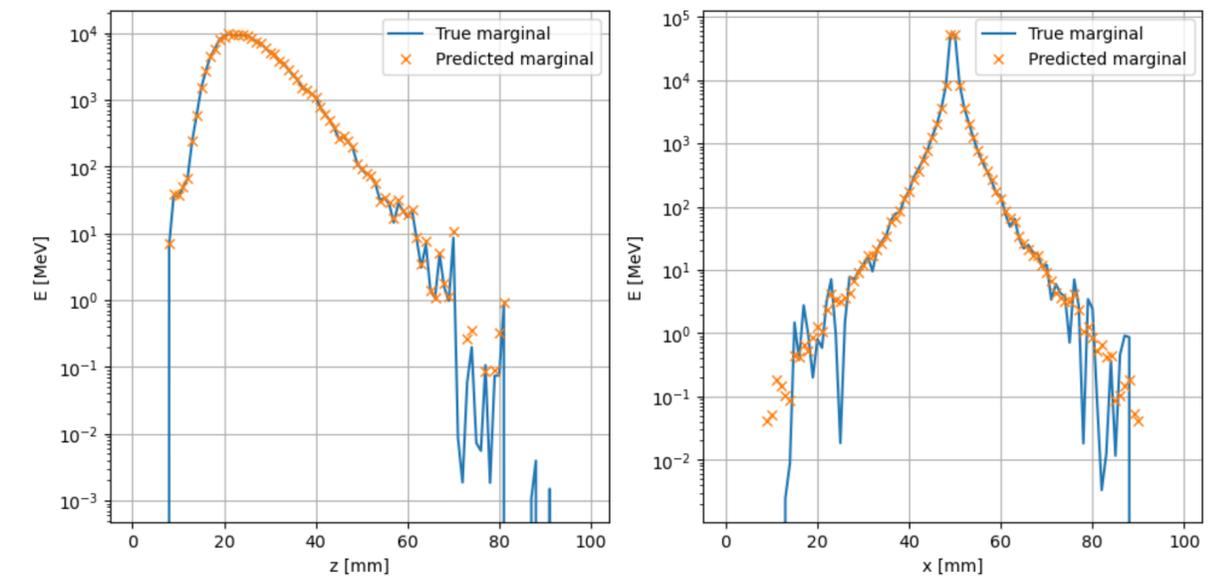
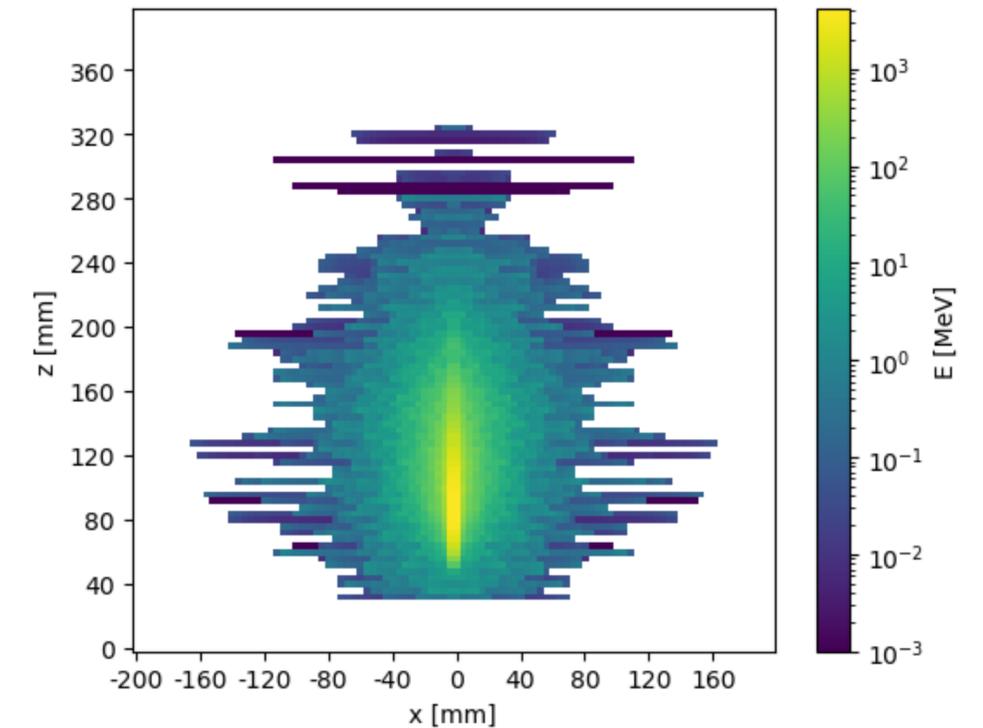
- Use the dataset to train a simple GravNet-based GNN to generate photon showers given an input energy.
- Not extremely precise, but
  - Fast
  - Describes well enough shower bulk
  - Total deposit matches the target
- To be patched with conditional DDPM (Denoising Diffusion Models). See Xuan Tung's talk at this year's MODE workshop ( <https://indico.cern.ch/event/1481852/contributions/6464894/> )



# Modules

## Shower Generation

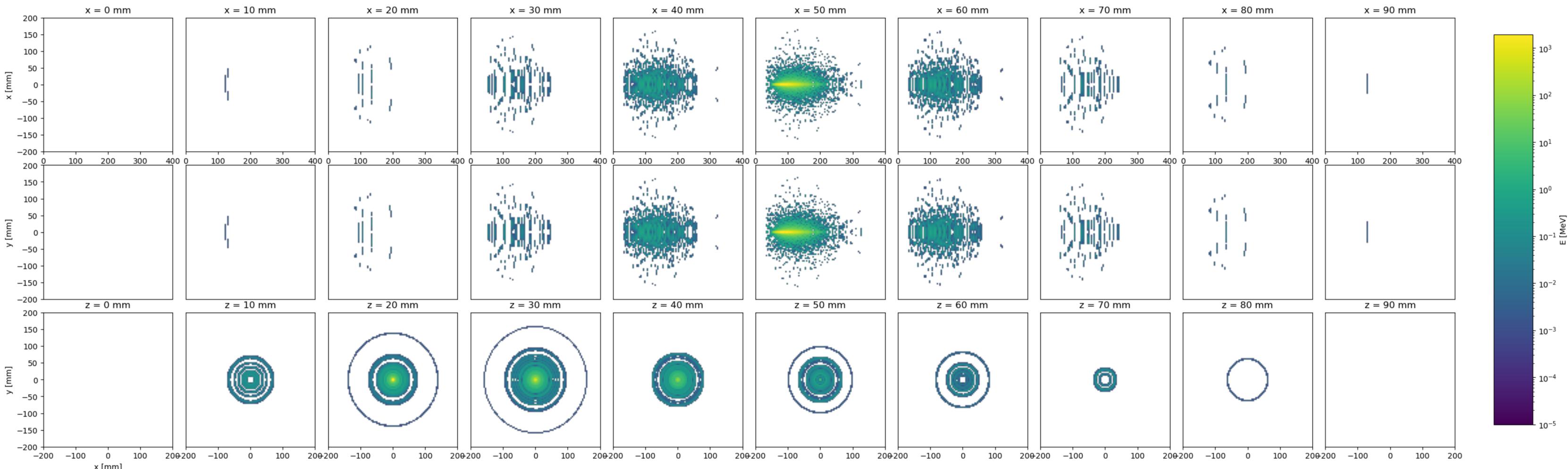
- To reconstruct the 3D shower shape from a marginal distribution we set up a minimal chi2 regressor
- Enforce cylindrical symmetry by introducing a numerical Jacobian modulating inferred 3D deposits
- X- and Z-marginals match the original distribution



# Modules

## Shower Generation

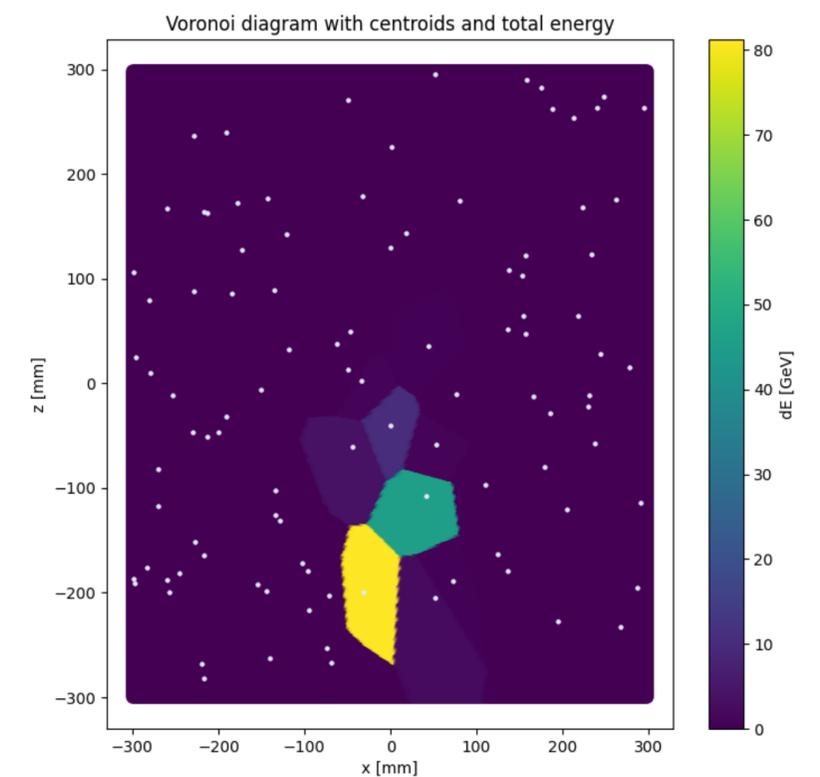
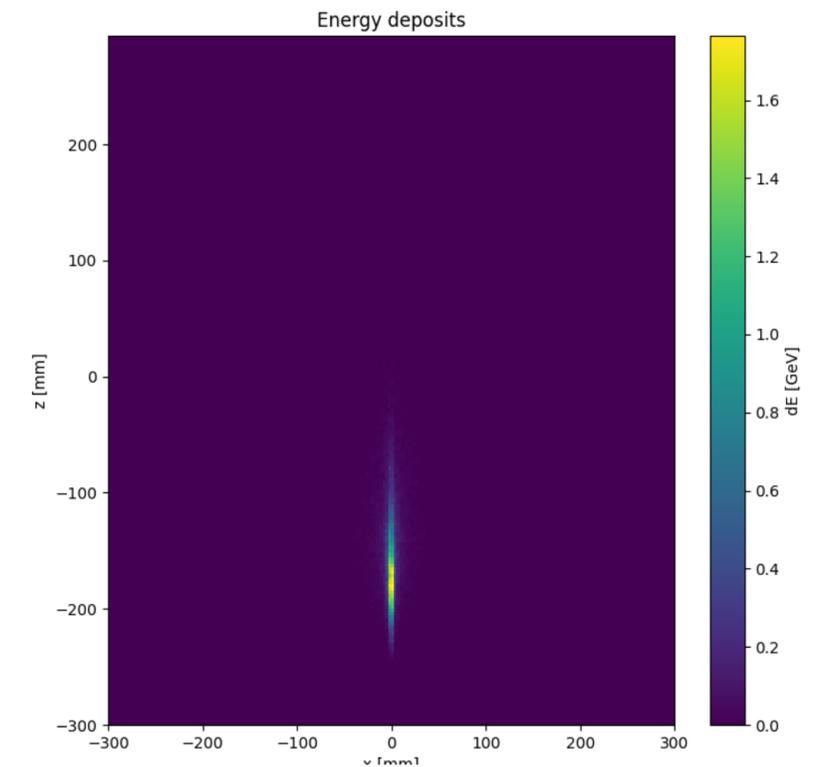
- Looking at slices of the simulated volume, shower features and symmetries are respected
- Cylindrical symmetry is correctly restored



# Modules

## Voxels and Event overlay

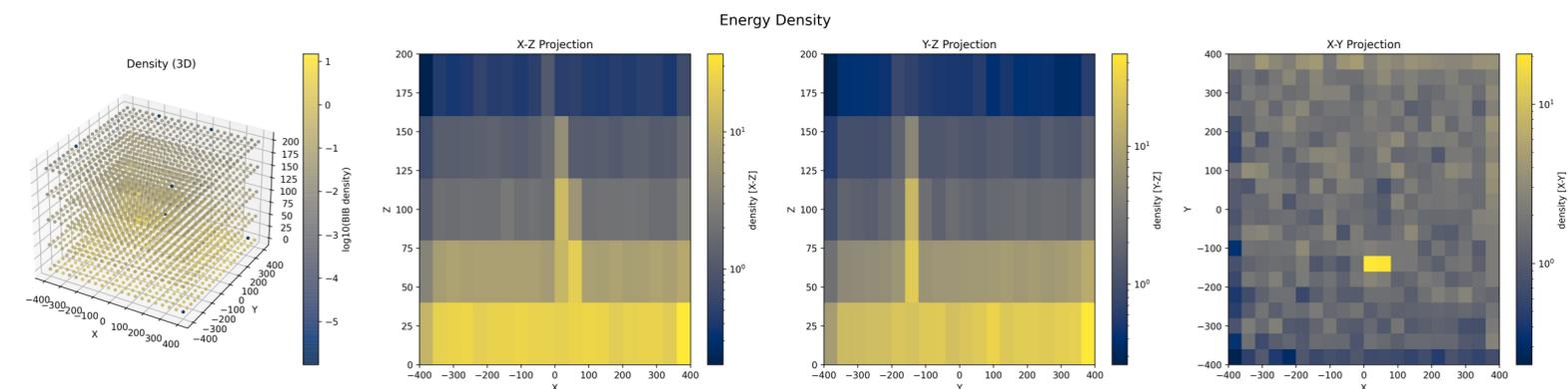
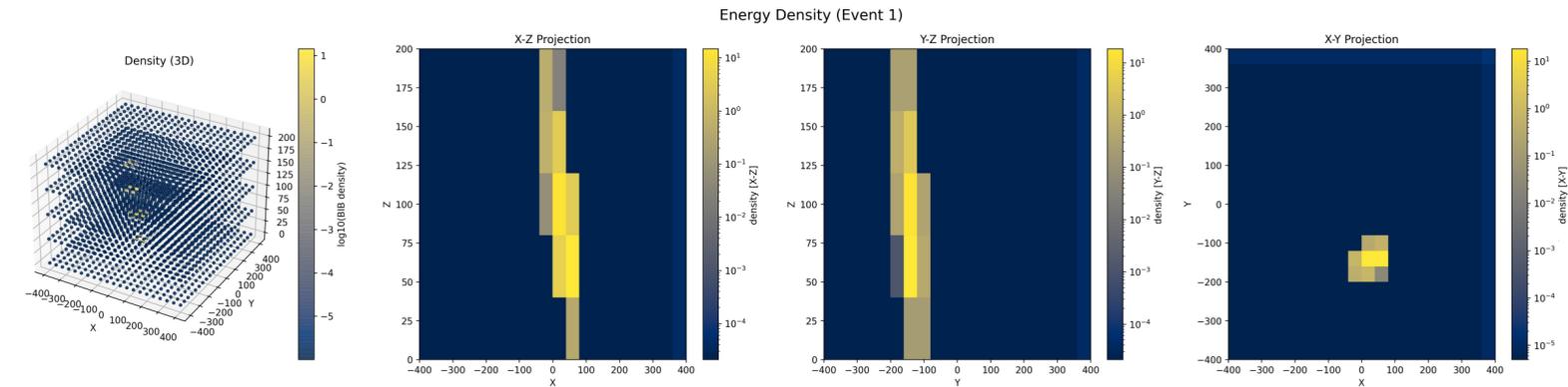
- Voxel structure is encoded through Voronoi regions identified by the voxel centers.
  - Boundaries defined as median planes between centers
- Objective is using centers as free parameters and thus optimize voxel location and extension
- In a  $(800 \times 800 \times 200) \text{mm}^3$  volume



# Modules

## Voxels and Event overlay

- 3D generated shower is randomized and assigned to Voronoi voxel
- Rotated:  $\phi \sim [0, 30] \text{deg}$   $\theta \sim [0, 360] \text{deg}$
- Entrance point randomly selected in a 20cm cube of base plane
- BIB density evaluated at voxel center, and scaled by voxel volume

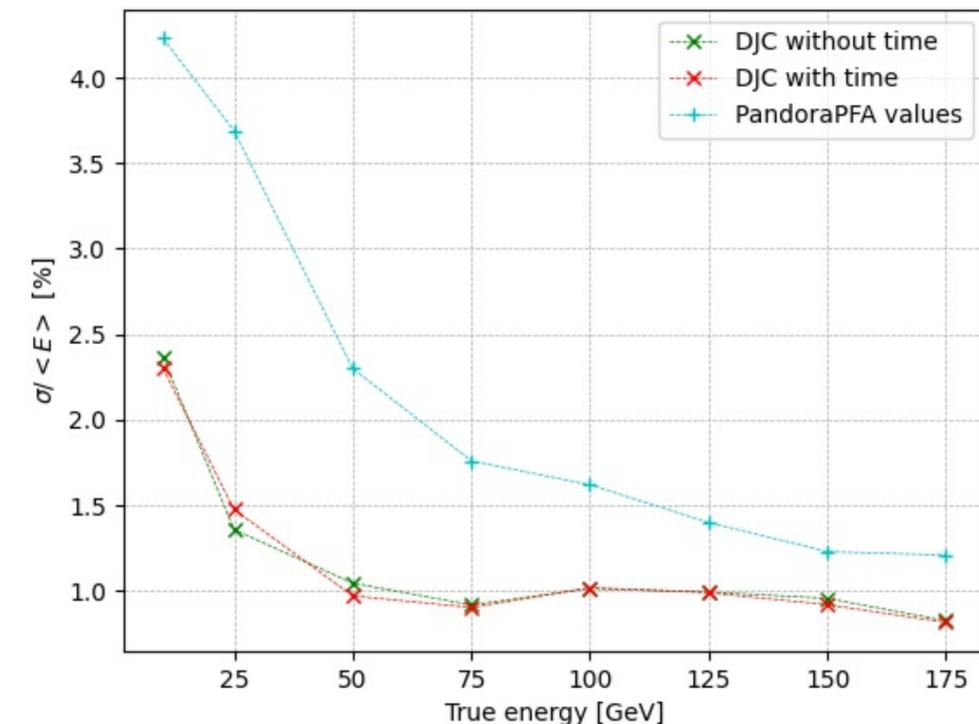


# Modules

## Reconstruction

<https://arxiv.org/abs/2204.01681>

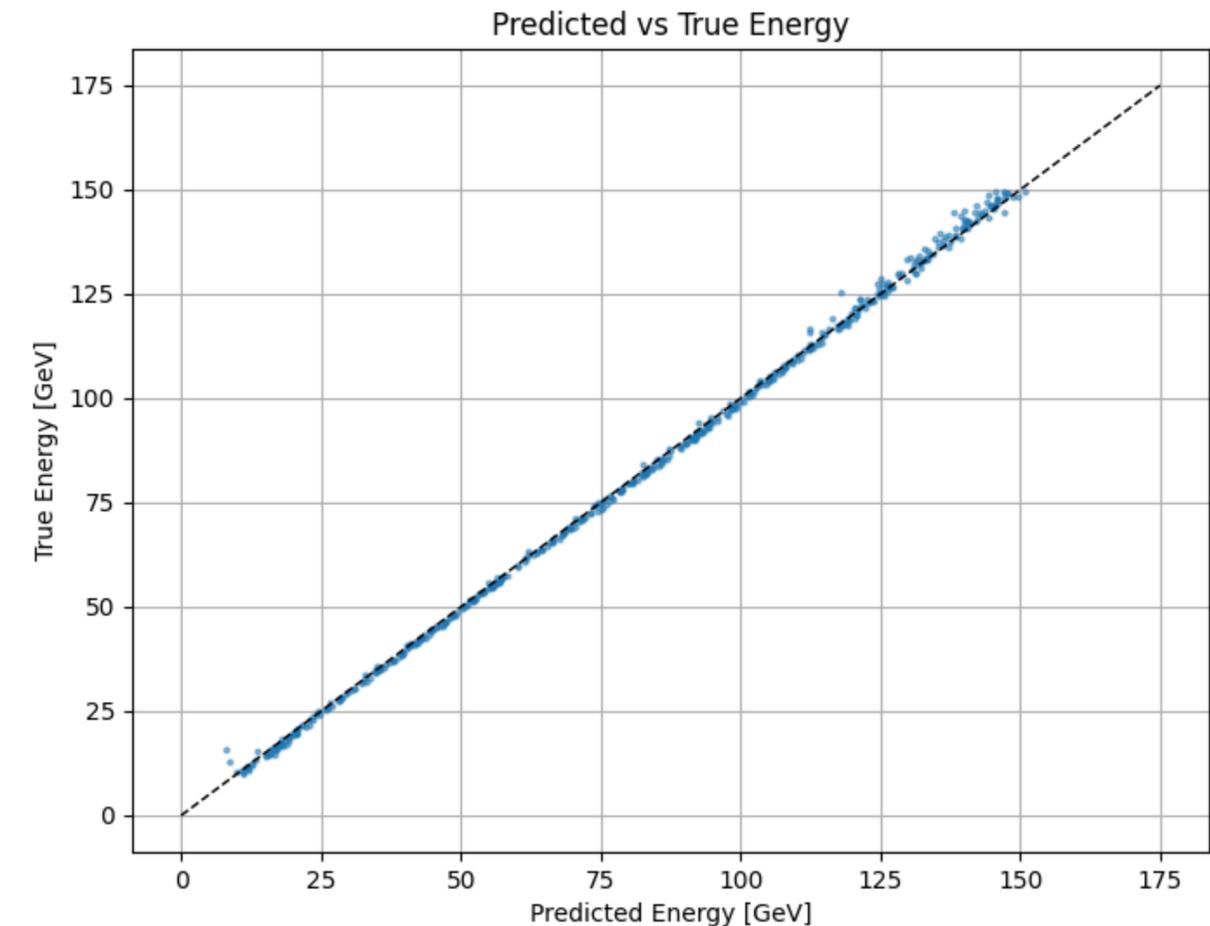
- Employed DeepJetCore for object reconstruction:
  - Signal photon vs BIB discrimination
  - Trained on 10k photons uniformly distributed in [10, 175]GeV
  - Tested on 8 fixed energy points
- Both signal and BIB overlaid from Geant4 simulations
  - Similar performance wrt collaboration framework on full BIB dataset
  - Significant improvement if we apply the same time window cut [-250,250]ps



# Modules

## Reconstruction

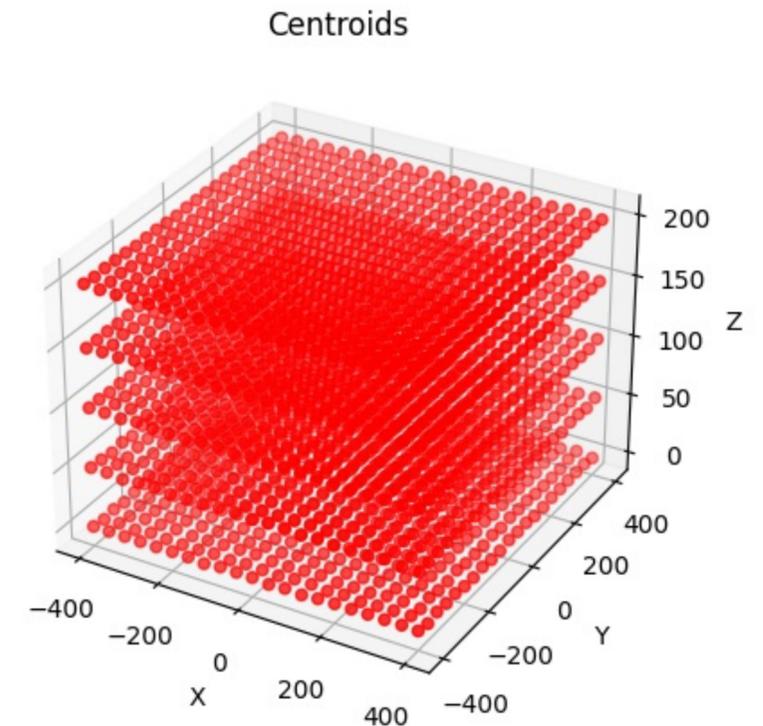
- With our framework clustering is trivial (only one cluster by definition)
- Kept the same architecture, adapted to simply regress a signal fraction associated to each cell deposit
- Trained on 10k events, generated with the developed surrogates on randomized centroid configuration (gaussian sampling centered at regular grid points, std the original cell dimension)
- Very good reconstruction performance, stable throughout different configurations



# Testing the Pipeline

## First run setup

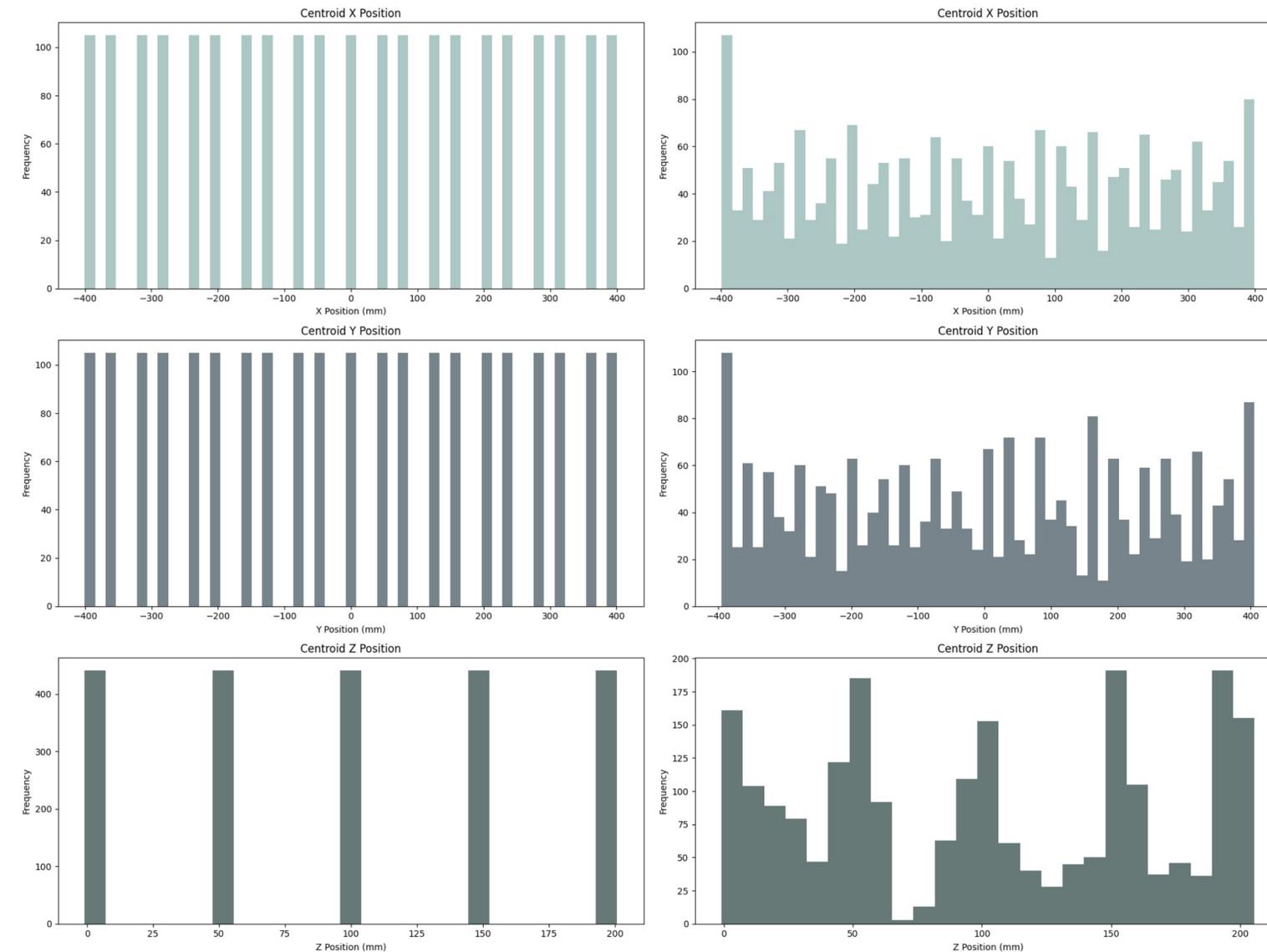
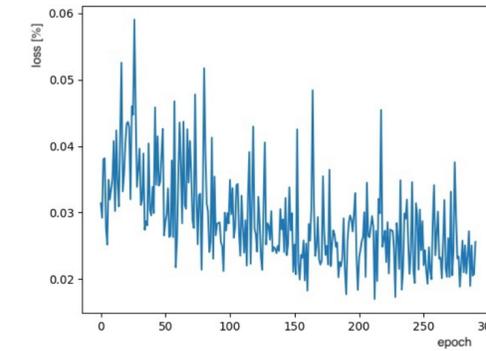
- Defined a basic utility function as energy reconstruction resolution  $\mathcal{L} := \frac{|E_{pred} - E_{true}|}{E_{true}}$
- Used initial configuration that mocks the original calorimeter design (4x4x5)cm cells in a 40x40x20cm<sup>3</sup> volume (Limited by GPU memory)
- Launched an optimization cycle over 300 epochs with linearly decaying learning rate (1.->0.1)



# Testing the pipeline

## First run Results

- Good news: pipeline works, gradients flow through and loss diminishes
- Still extremely noisy: Reconstruction solid enough to properly reconstruct photons, although we have already a small gain
- Can draw first considerations on how voxels diffuse

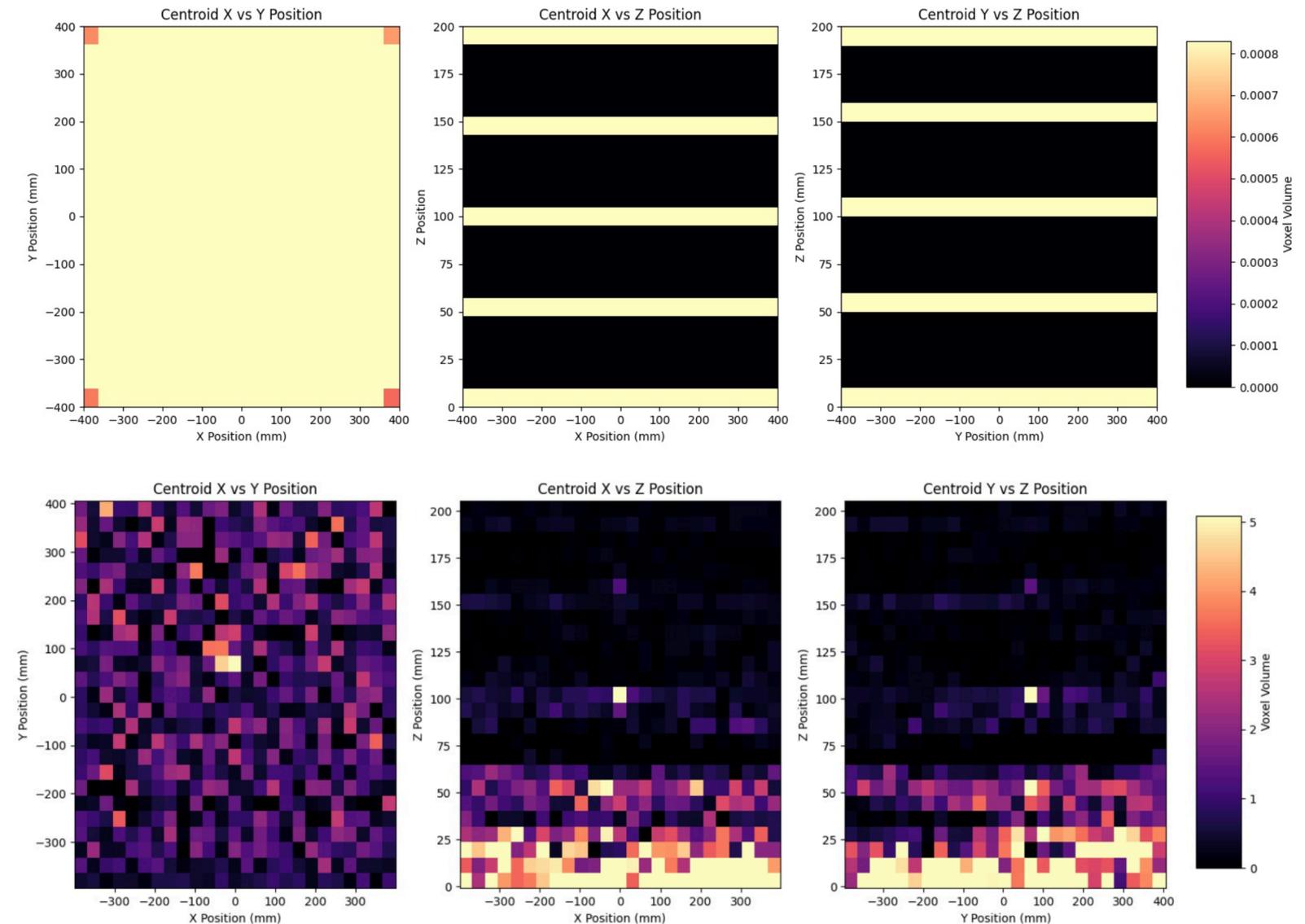


# Testing the pipeline

## First run Results

- Looking at volume maps, voxels on the bottom block tend to diffuse upwards
- Left with less, broader cells in the lower region
- There most information is lost due to BIB. Lower cells can play also as absorbers, increasing granularity on the layers above to collect information where signal is clearer.

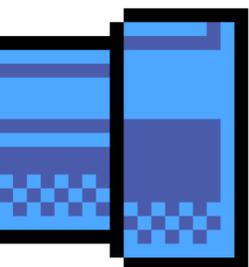
Note: not same scale, max of top plot is normalized to 1



# Wrap up

## Summary and conclusions

- Presented a **proof-of-concept** for a SGD-based optimization pipeline. Results are still preliminary, but provide a starting point for more realistic studies
- **Surrogate** models provide fast, differentiable tools for event simulation
- Solid reconstruction performance, allows for more complicated utility definition. Explore multi-target functions, discussing with collaboration and encoding **geometrical** limitations and preferences. Implementation of **position** reconstruction as well.



**Thank you!**

**Backup**

# 2D image to 3D shower

## Chi2 regressor

---

**Algorithm 2** Reconstruction of 3D Energy Distribution from a Marginal using Cylindrical Symmetry

---

**Input:** Target marginal image  $f_Y(x,z) \in \mathbb{R}^{n \times m}$

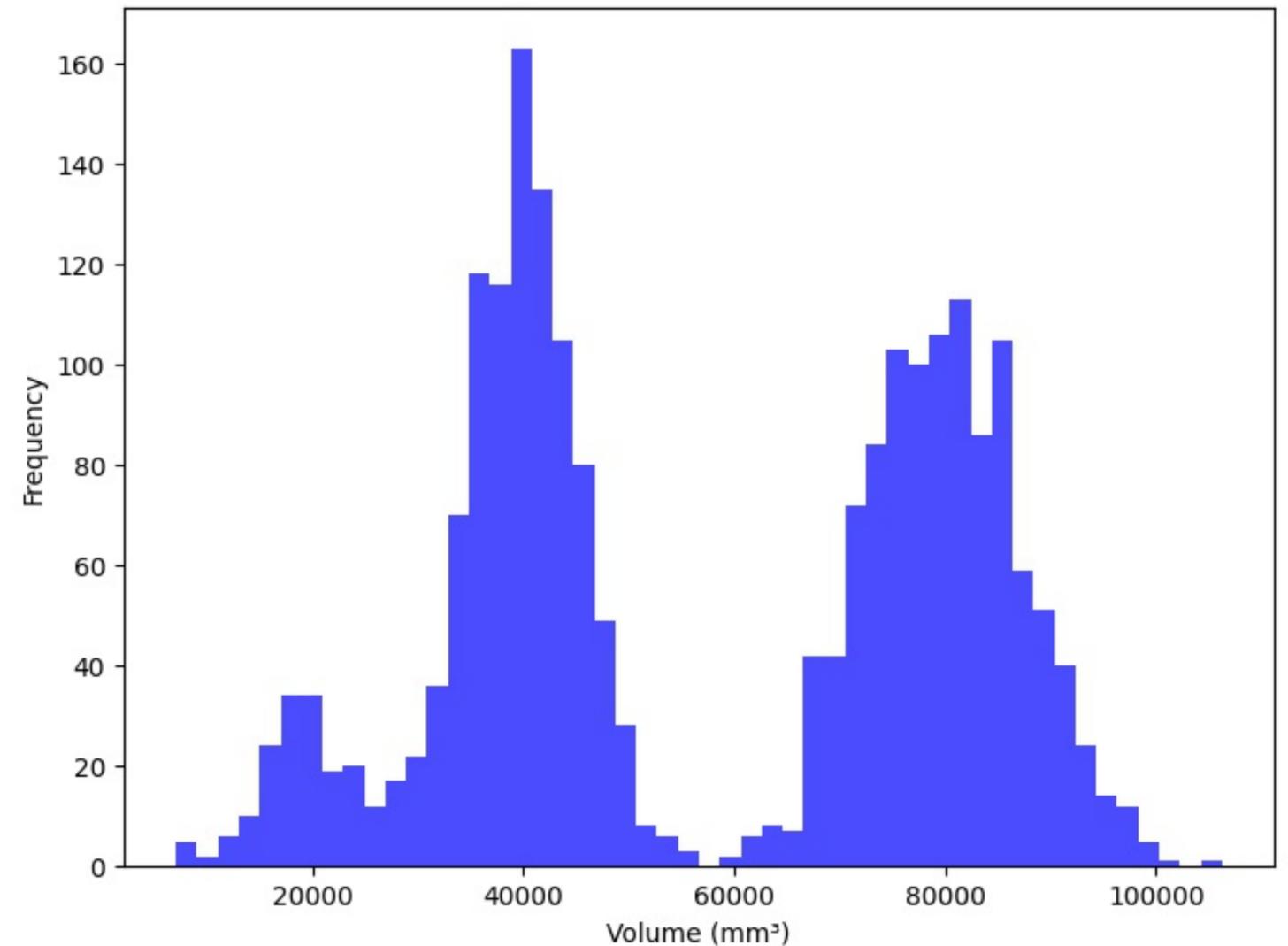
**Output:** Predicted 3D distribution  $E(x,y,z) \in \mathbb{R}^{n \times m \times p}$

- 1: **procedure** InflateShower( $f_Y$ )
  - 2:   Precompute cylindrical ring fractions  $f$  via uniform sampling
  - 3:   Initialize density tensor  $\rho$  as trainable variable
  - 4:   **for**  $t = 1$  to  $T$  (training steps) **do**
  - 5:     Compute predicted marginal:  $\hat{f}_{Y_{i,j}} = \sum_k \sum_m \sum_l f_{ijkl} \rho_{lm}$
  - 6:     Compute loss:  $L = \sum_{ij} (\hat{f}_{Y_{ij}} - I_{ij})^2$ .
  - 7:     Update  $\rho$  using Adam optimizer and gradients from  $L$
  - 8:     Enforce physicality:  $\rho \leftarrow \max\{\rho, 0\}$
  - 9:   **end for**
  - 10:   Reconstruct full 3D energy volume:  $E_{ijk} = \sum_m \sum_l f_{ijkl} \rho_{lm}$
  - 11:   **return**  $E$
  - 12: **end procedure**
-

# Voxel volume estimation

## Monte-Carlo sampling for Voronoi volumes

- Voxel volumes are estimated through Monte Carlo sampling
- Draw 200k points over detection volume
- Assign each to the closest voxel center
- At initial configuration 3 type of voxels, depending on whether they occupy a central, edge or corner position



# Muon Collider details

## Muon production beamline

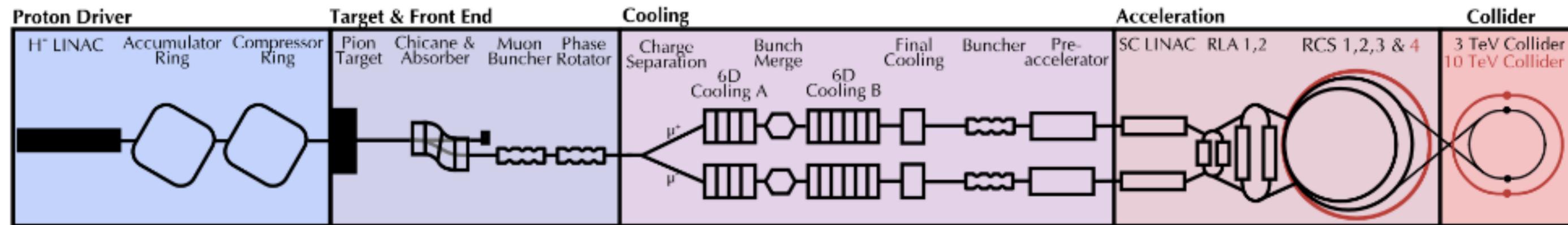


Figure 3.2: Schematic layout of the Muon Collider system. From [96]

# Muon Collider details

## Envisioned timeline

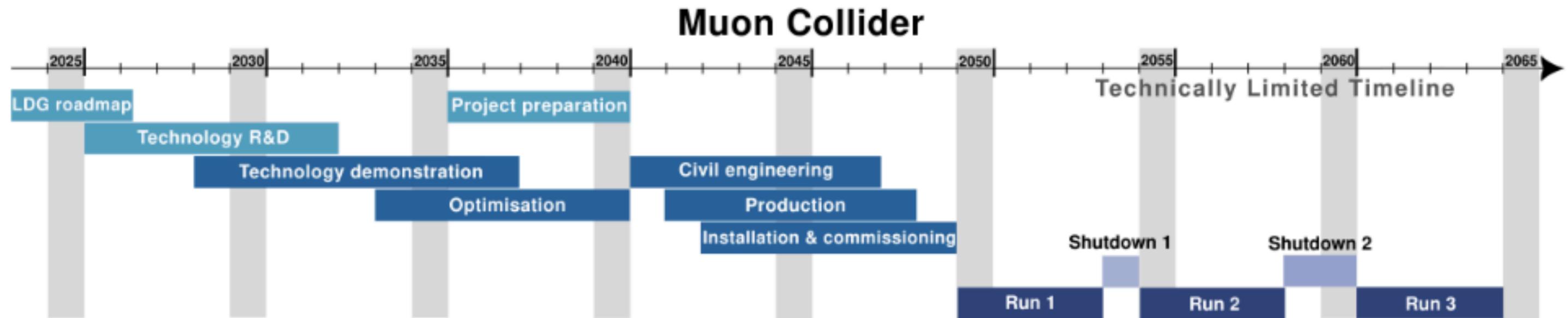


Figure 3.1: Proposed R&D and construction milestones needed to enable a first 3 TeV stage by 2050, assuming a successful demonstration of cooling, magnets, and detectors

# Muon Collider details

## Target parameters

- Scenario 1: Energy staging. First run at 3TeV, there are physics case studies, lower instrumentation cost at beginning.
- Scenario 2: luminosity staging. Less performing magnets, but already at full energy. Full cost required at beginning. Estimated factor 3 loss in luminosity

Parameter	Symbol	Unit	Scenario 1		Scenario 2	
			Stage 1	Stage 2	Stage 1	Stage 2
Center-of-mass energy	$E_{cm}$	TeV	3	10	10	10
Target integrated luminosity	$\int \mathcal{L}_{target}$	$\text{ab}^{-1}$	1	10	10	10
Estimated luminosity	$\mathcal{L}_{estimated}$	$10^{34} \text{cm}^{-2} \text{s}^{-1}$	2.1	21	5(tbc)	14
Collider circumference	$C_{coll}$	km	4.5	10	15	15
Collider arc peak field	$B_{arc}$	T	11	16	11	11
Luminosity lifetime	$N_{turn}$	turns	1039	1558	1040	1040
Muons/bunch	$N$	$10^{12}$	2.2	1.8	1.8	1.8
Repetition rate	$f_r$	Hz	5	5	5	5
Beam power	$P_{coll}$	MW	5.3	14.4	14.4	14.4
RMS longitudinal emittance	$\epsilon_{\parallel}$	eV	0.025	0.025	0.025	0.025
RMS transverse emittance	$\epsilon_{\perp}$	$\mu\text{m}$	25	25	25	25
IP bunch length	$\sigma_z$	mm	5	1.5	tbc	1.5
IP beta function	$\beta$	mm	5	1.5	tbc	1.5
IP beam stipe	$\sigma$	$\mu\text{m}$	3	0.9	tbc	0.9
Protons on target/bunch	$N_p$	$10^{14}$	5	5	5	5
Proton energy in target	$E_p$	GeV	5	5	5	5

Table 3.1: Tentative target parameters for a Muon Collider at different energies.