

CaloClouds: Fast Geometry-Independent Highly-Granular Calorimeter Simulation

CaloChallenge Workshop

Material based on: [arXiv:2305.04847v1](https://arxiv.org/abs/2305.04847v1)

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Katja Krüger, Peter McKeown, Lennart Rustige

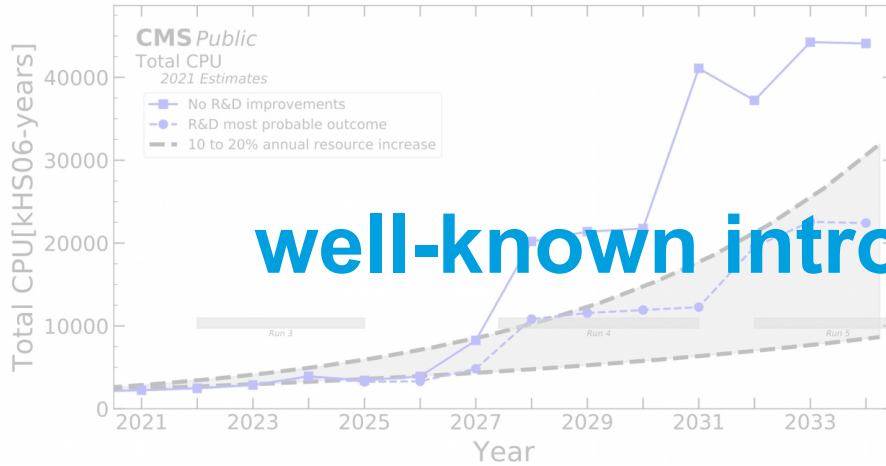
¹ **Deutsches Elektronen-Synchrotron**, anatolii.korol@desy.de

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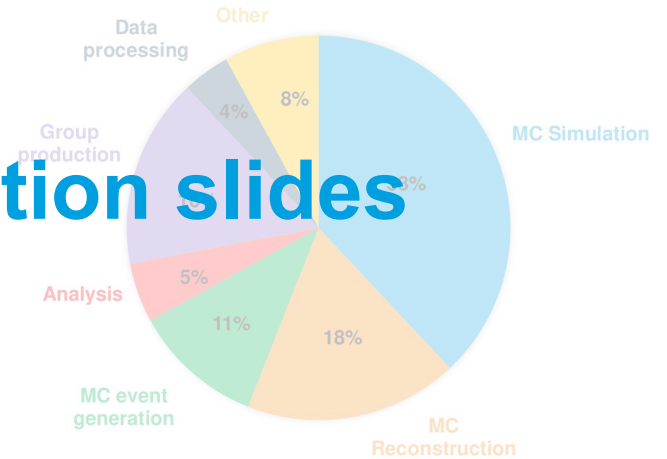


Problem Definition

Time-consuming Simulations



WALL CLOCK CONSUMPTION PER WORKFLOW



D. Costanzo, J. Catmore, ATLAS
Computing update, LHCC meeting , 2019

Goal: replace (or augment) simulation steps with a faster powerful generator, based on state-of-the-art machine learning techniques

Generative Models

Overview

- Generative Model is just a function that maps random numbers to some structure
- In most cases the structure is an **image representation** of the electromagnetic shower (EM shower) in the calorimeter

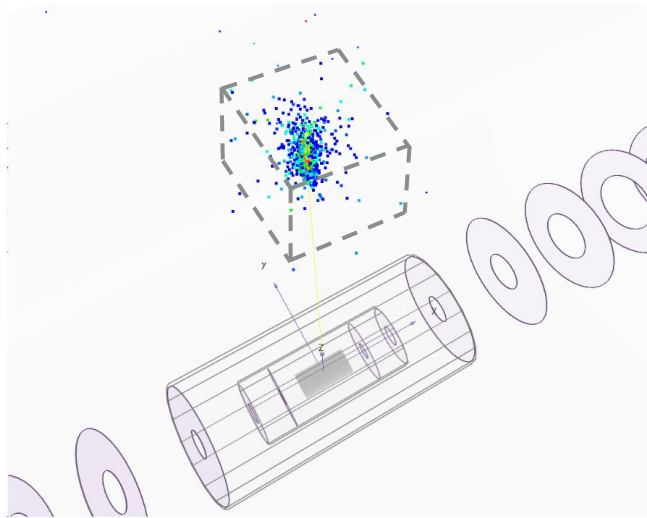


- There exist numerous generative models
 - Generative Adversarial Networks (GANs)
 - Autoencoders (AE), e.g. BiB-AE
 - Flow-based models
 - Denoising Diffusion Probabilistic models (DDPs)

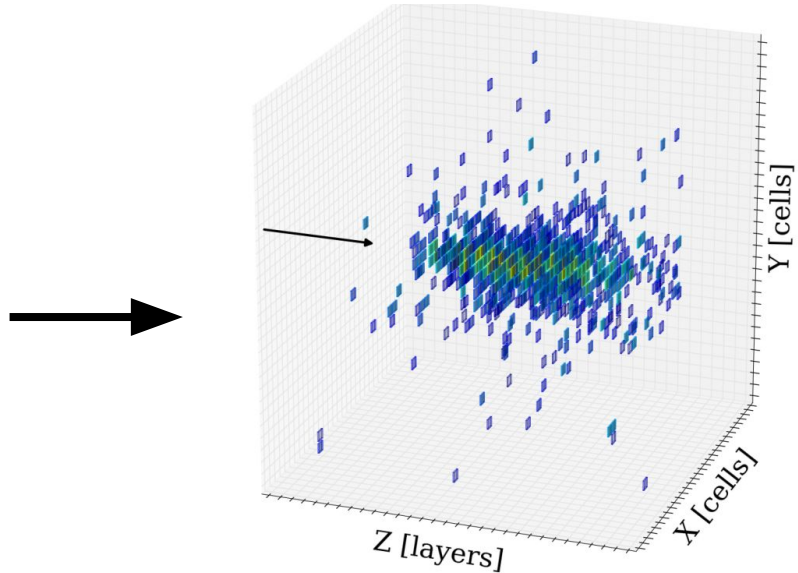
Image Representation of the EM Showers

ILD Detector

A simulated 60 GeV photon shower in the ILD detector



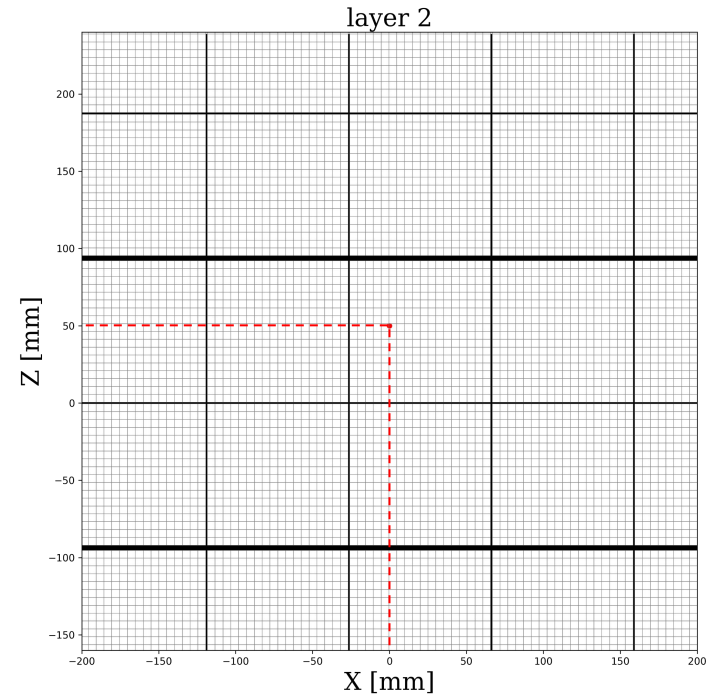
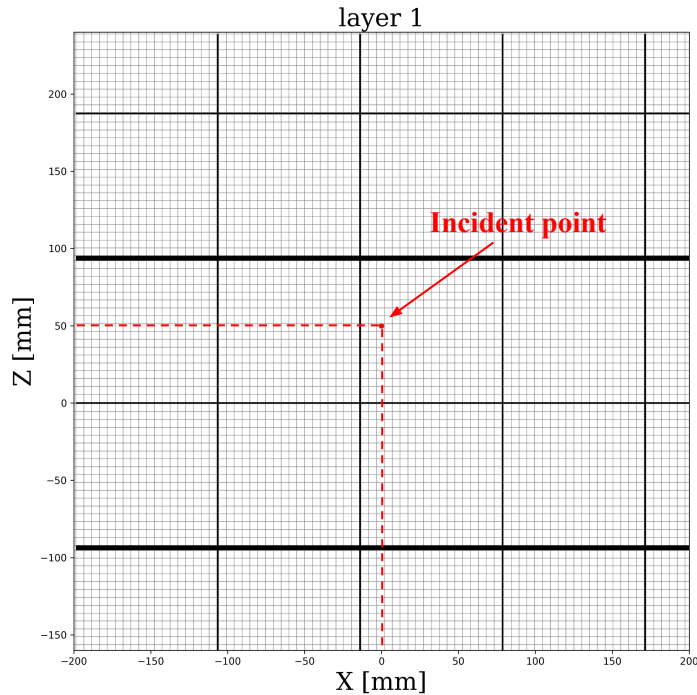
Regular grid 30x30x30



One to one mapping from detector geometry to a regular grid

Image Representation of the EM Showers

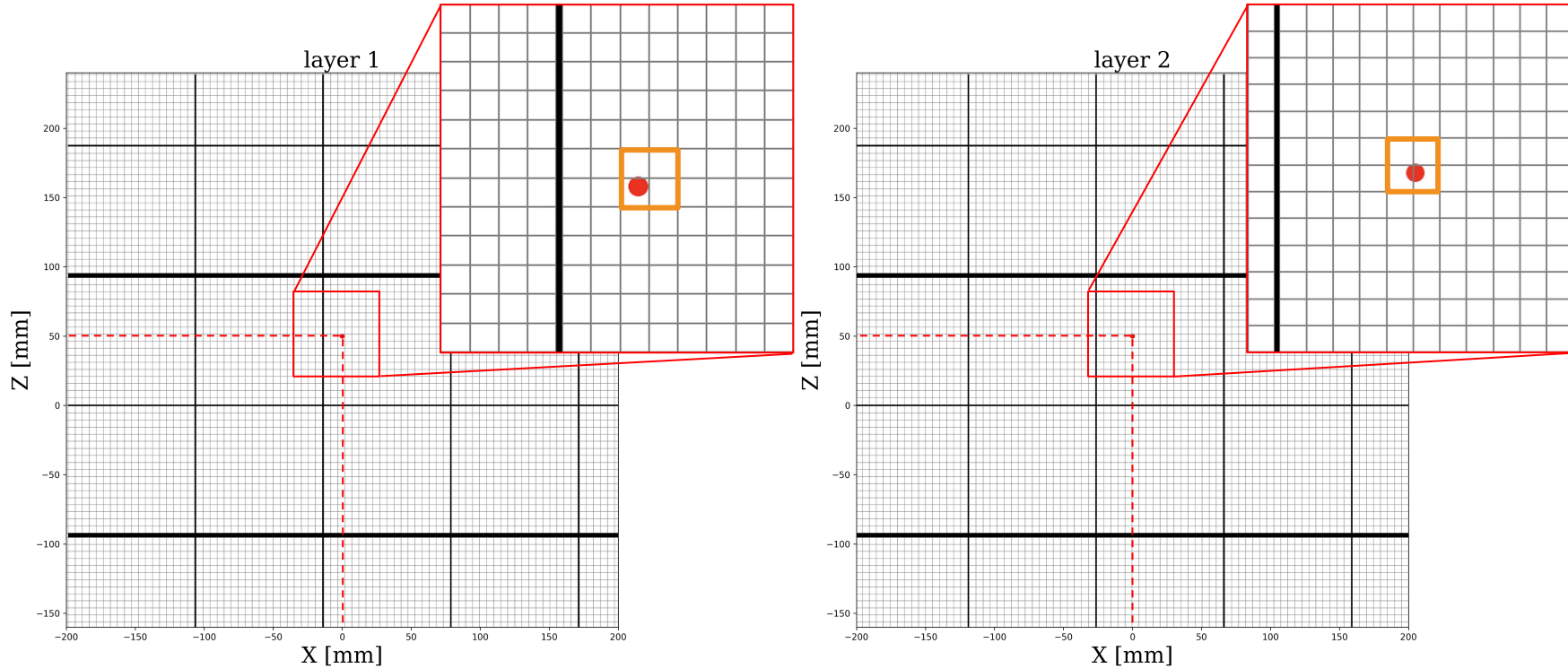
ILD Detector, ECAL Layers Structure



White squares represent **active cells**. Black lines are wafers, construction gaps, etc. (**not active material**)

Image Representation of the EM Showers

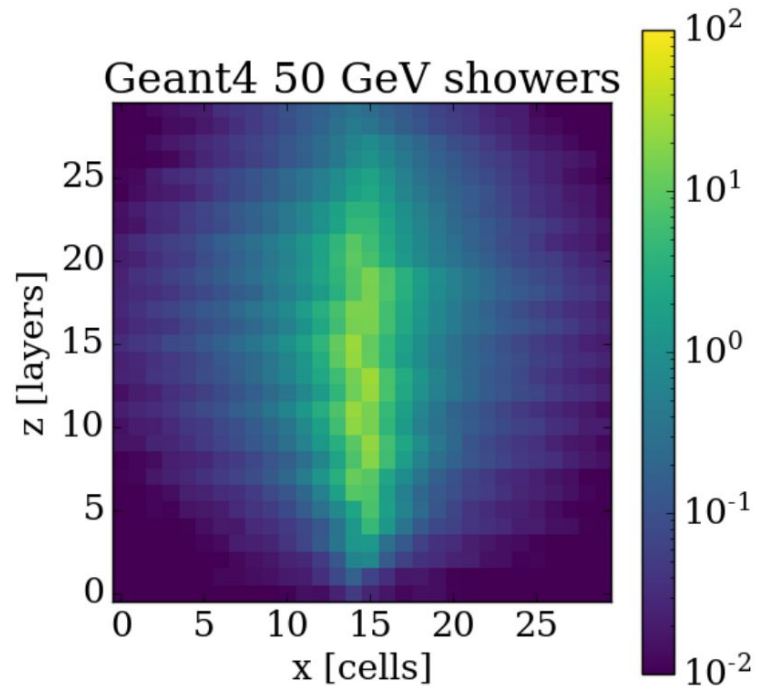
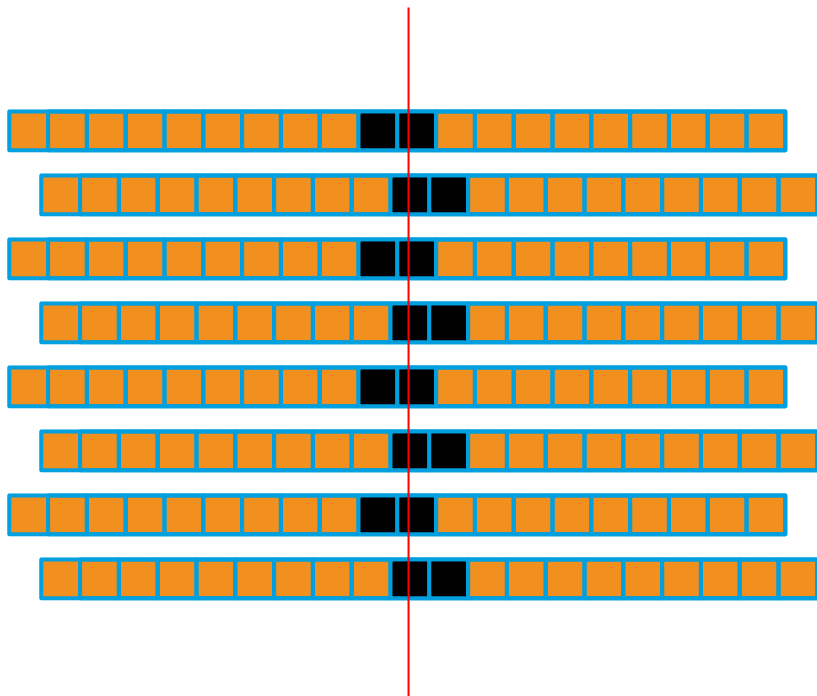
ILD Detector, ECAL Layers Structure



White squares represent **active cells**. Black lines are wafers, construction gaps, etc. (**not active material**)

Image Representation of the EM Showers

ILD Detector, ECAL Layers Structure, Staggering Effect



Models have to learn not only EM shower properties, but also geometry “artifacts”, like **staggering effect**

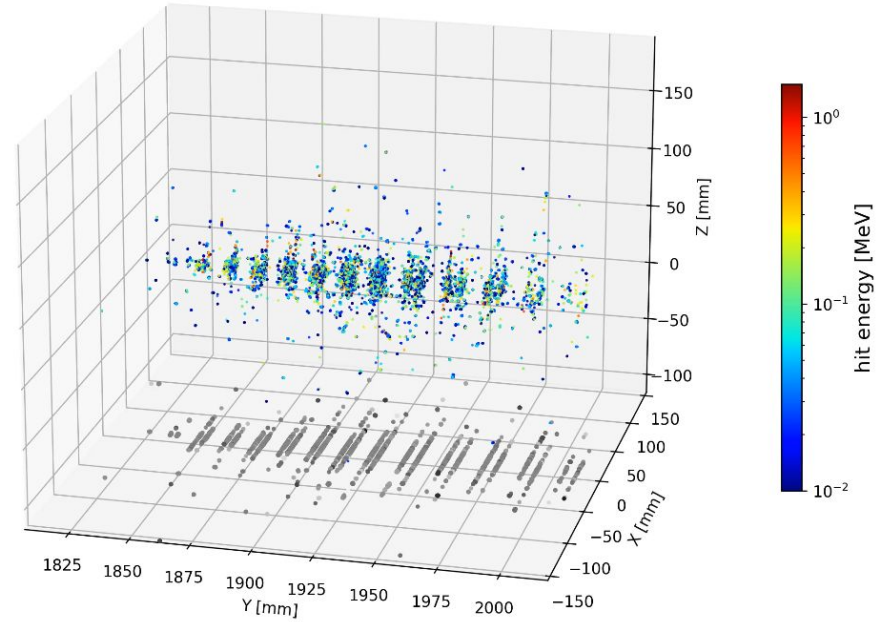
Point Cloud Representation of the EM Showers

GEANT4 Steps

A way to overcome potential issues from irregular (realistic) cell geometry would be to use much higher granularity/resolution

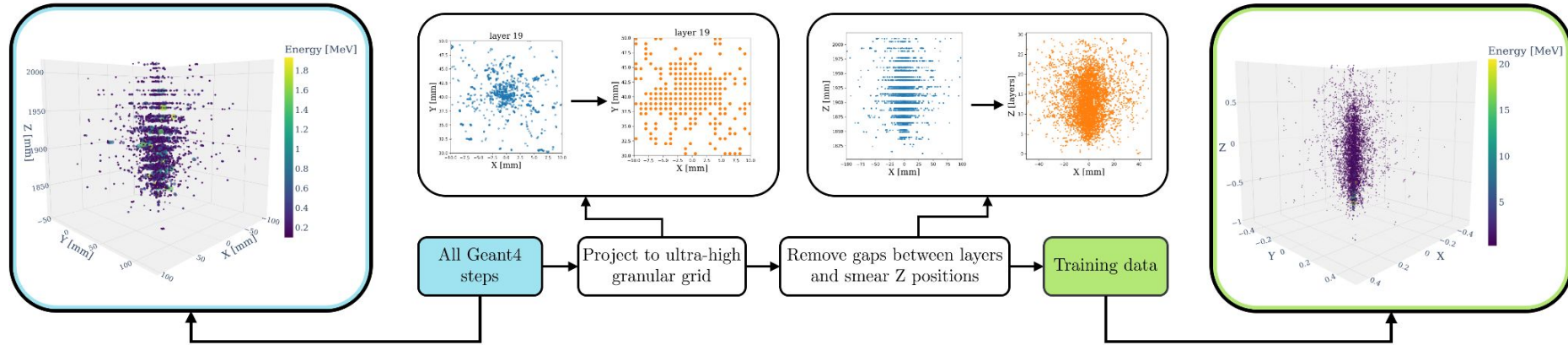
- All G4 interactions, ultimate resolution
- Detached from detector layer geometry
- Too many points to generate, $\sim 40k$ per shower (need pre-processing step to reduce number of spacepoints)

Photon
Energy: 90 [GeV]
Event: 4
Time step: 0.98246 [ns]



Point Cloud Representation of the EM Showers

Data Preprocessing

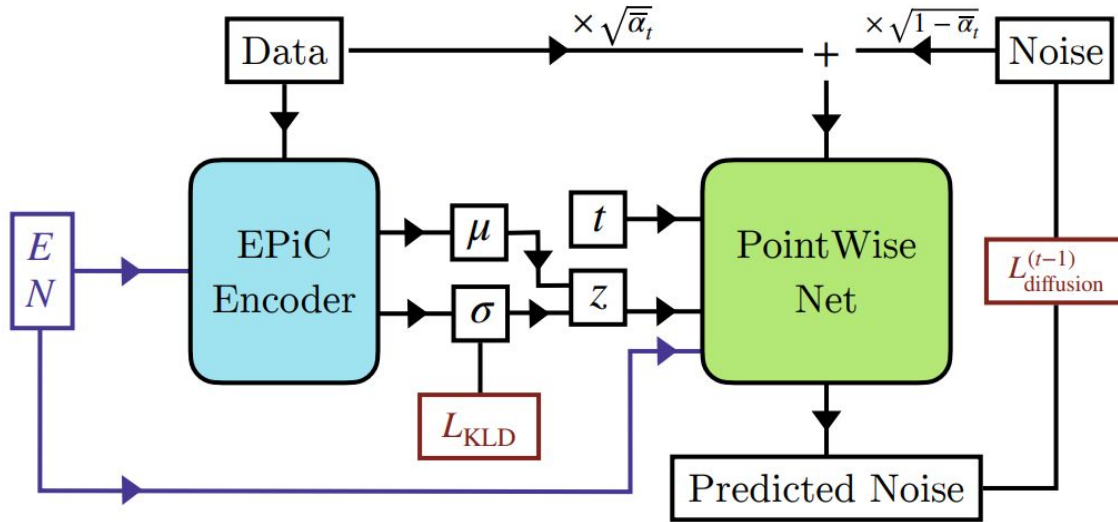


Number of points reduced to ~6k per shower, high enough resolution to move the shower in different place without harming physical properties of the shower

Point Cloud + Diffusion Model

Model Overview, Training

CaloClouds: Fast Geometry-Independent
Highly-Granular Calorimeter Simulation
[arXiv:2305.04847v1](https://arxiv.org/abs/2305.04847v1)



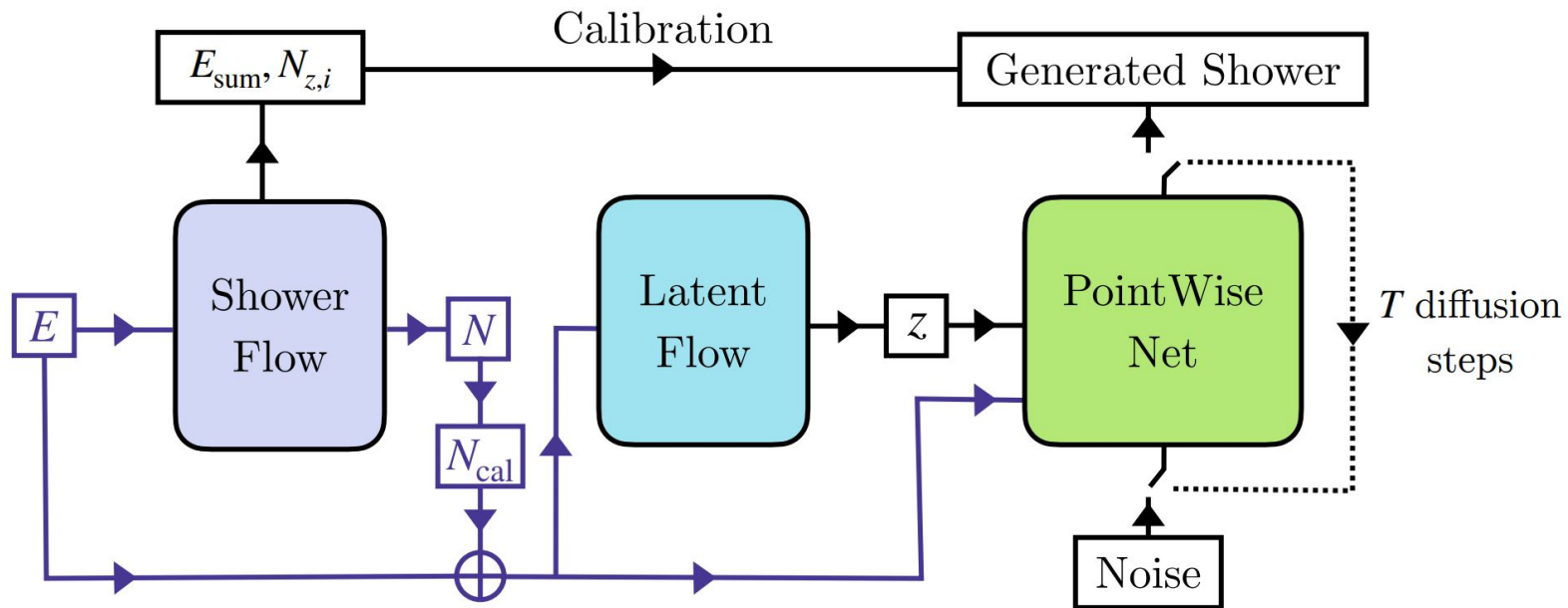
(a) Training at random time step t

- GANs and VAEs convert noise from some simple distribution to a data sample
- **DMs learn to gradually denoise data starting from noise**

Point Cloud + Diffusion Model

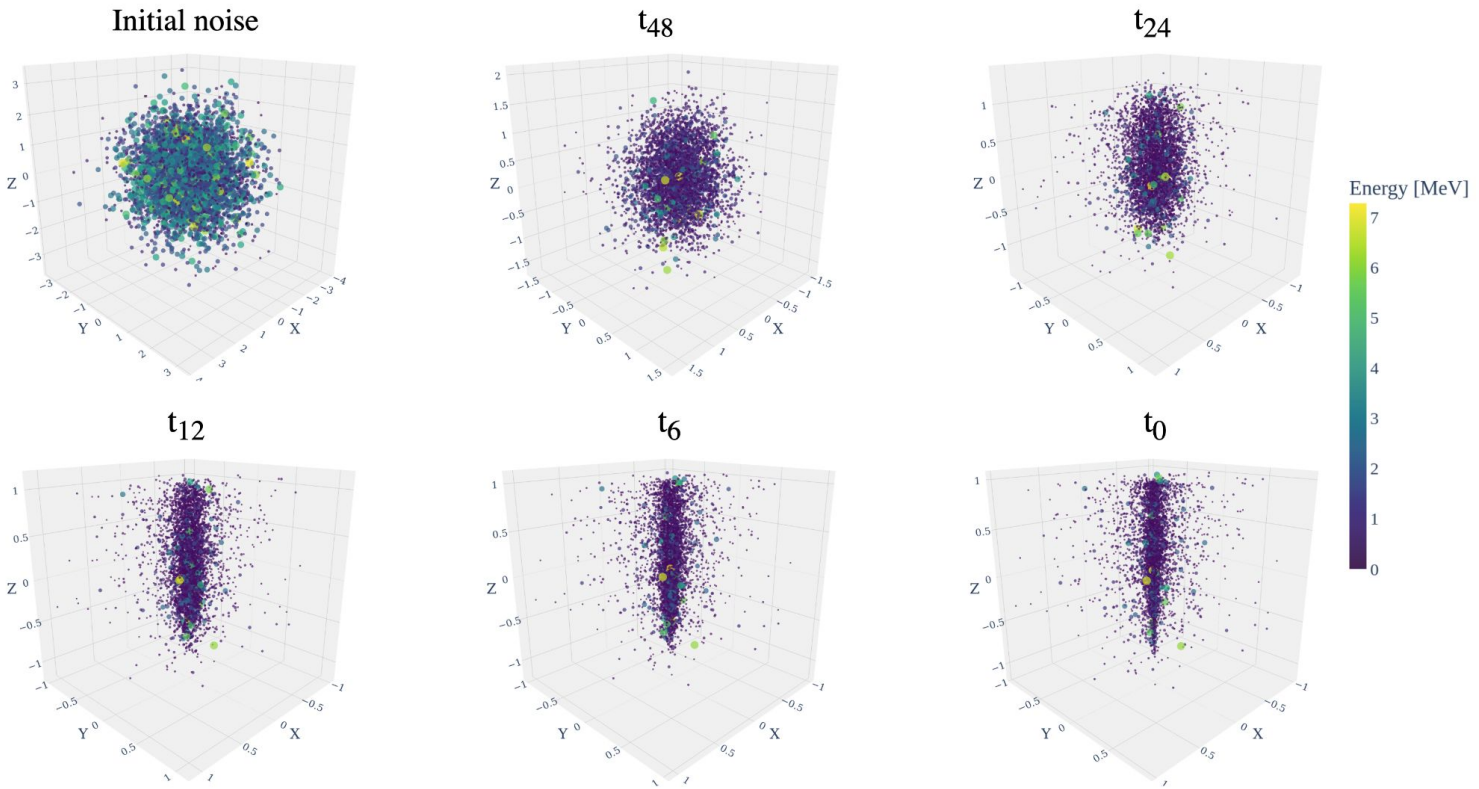
Model Overview, Inference

CaloClouds: Fast Geometry-Independent
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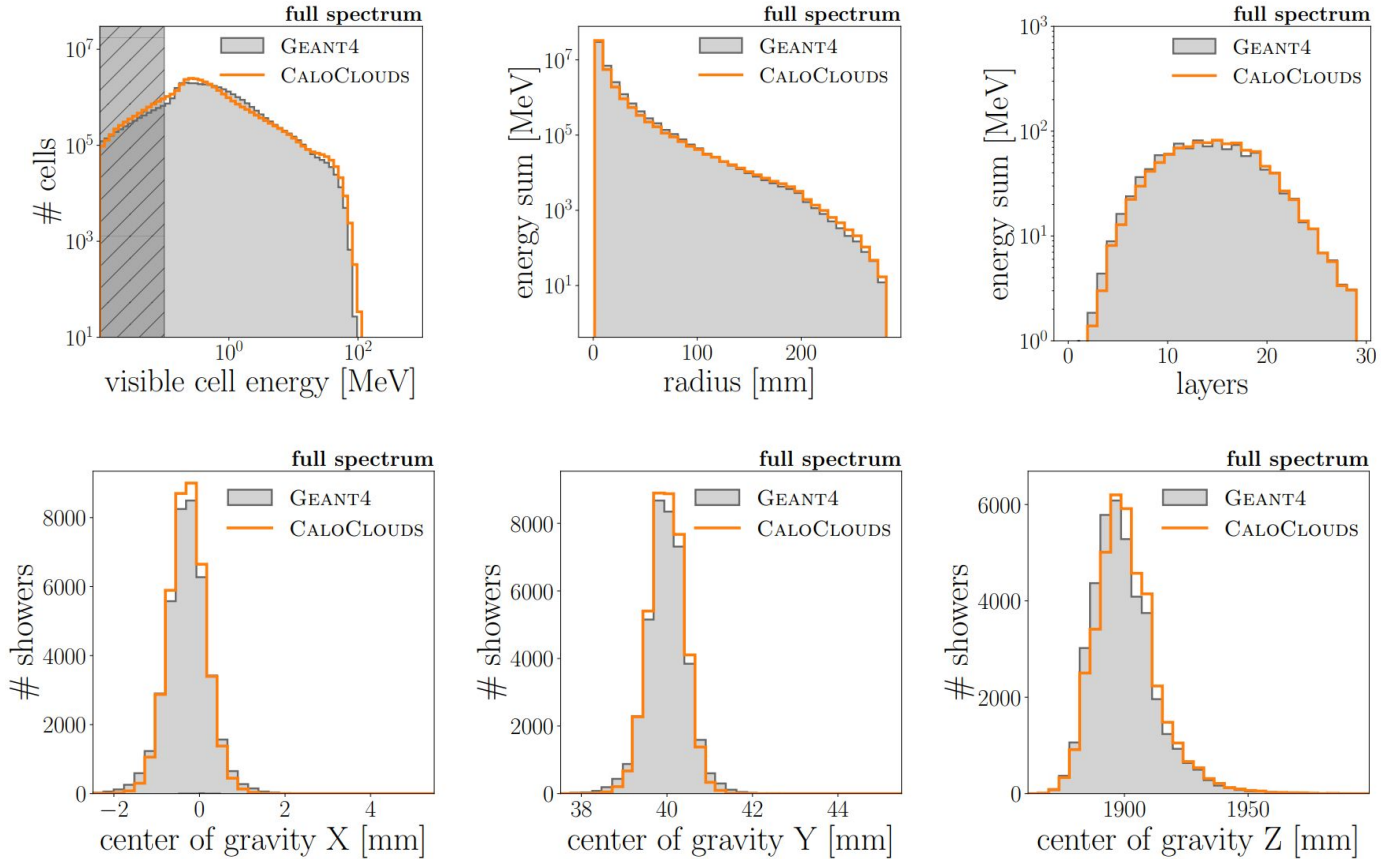
Point Cloud + Diffusion Model

Reverse Diffusion Process



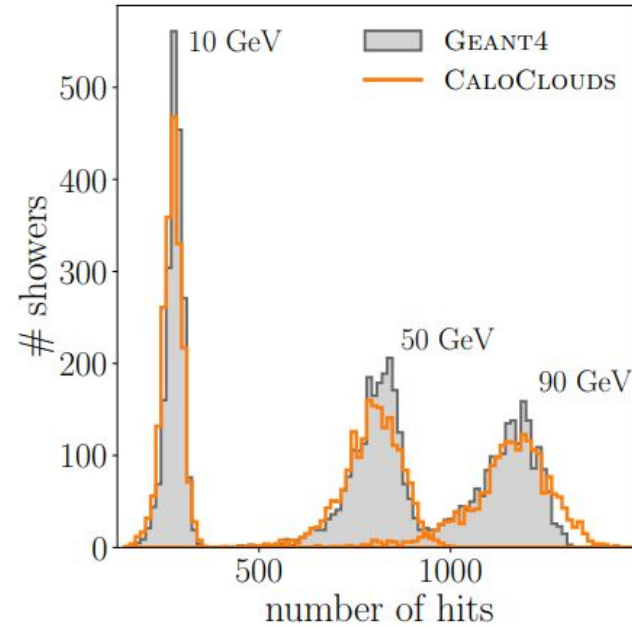
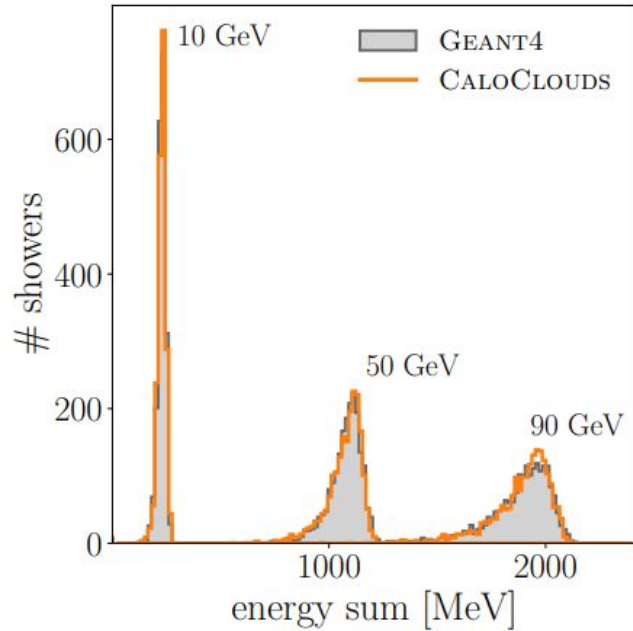
Point Cloud + Diffusion Model

Results



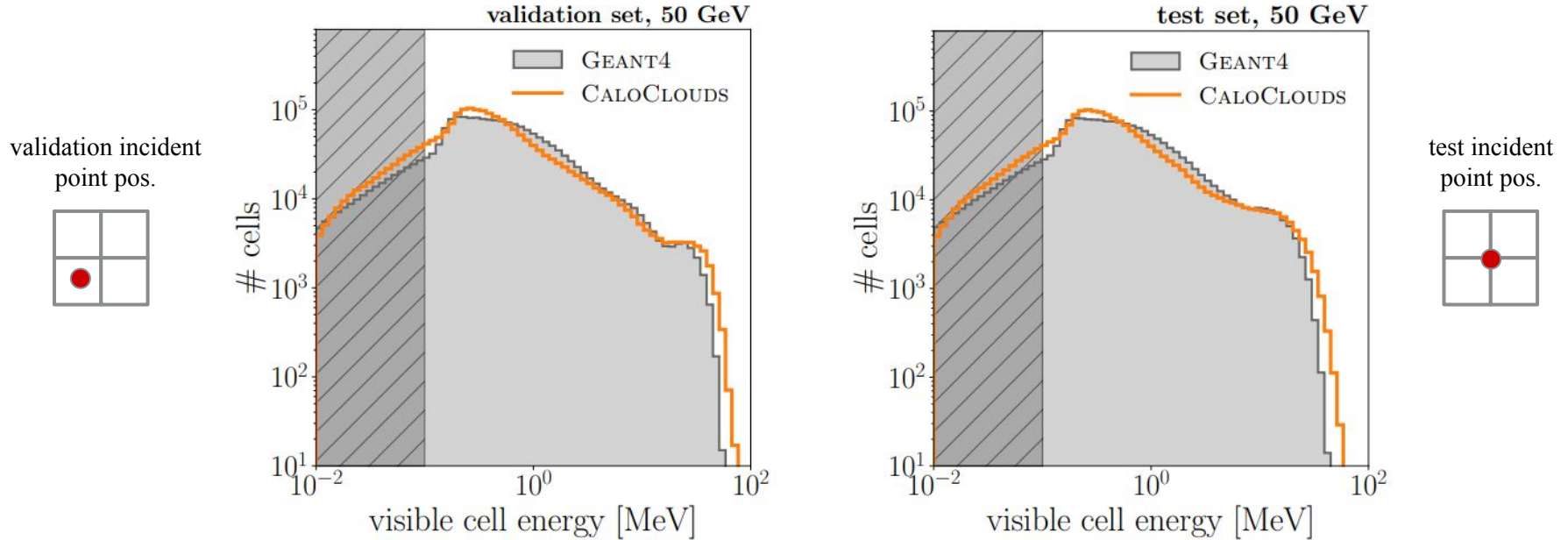
Point Cloud + Diffusion Model

Results



Point Cloud + Diffusion Model

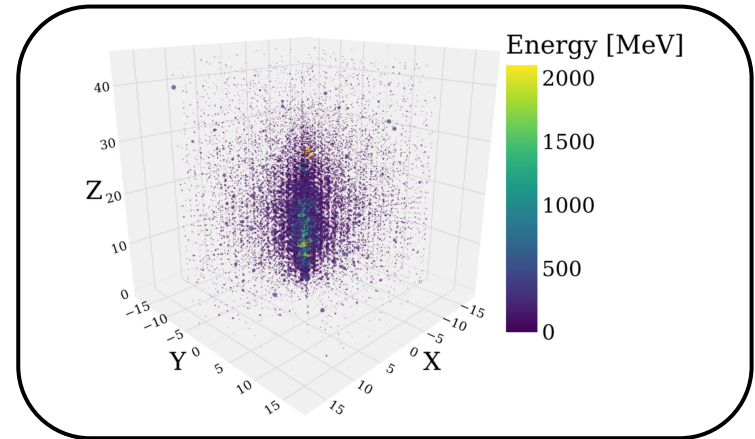
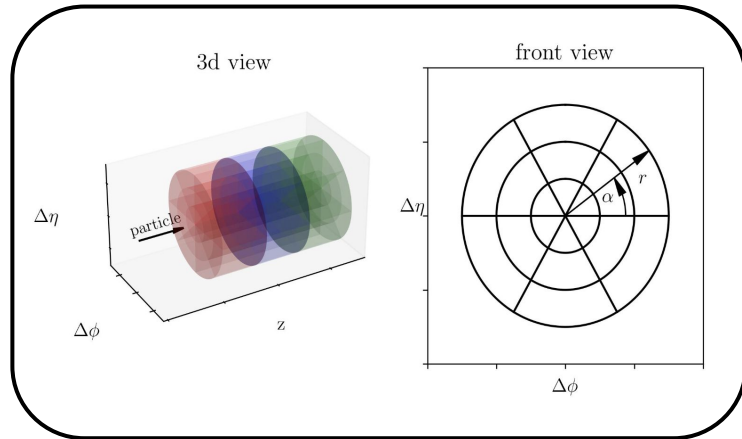
Results



Per-cell energy distribution for the 50 GeV validation (left) data set, created at the same position as the training data set and for a 50 GeV test (right) data set simulated at a different position with the generated point cloud translated to this position

CaloChallenge

Data Preprocessing



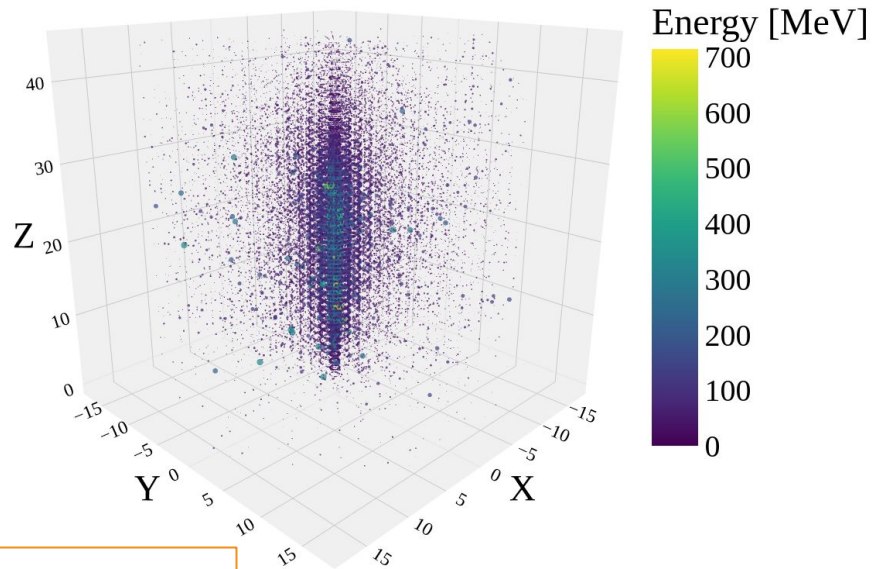
dataset 3, electron, 1 TeV

- Cylindrical to cartesian coordinates transform
- Incident energy normalization, feature scaling
- Coordinates normalization, feature scaling

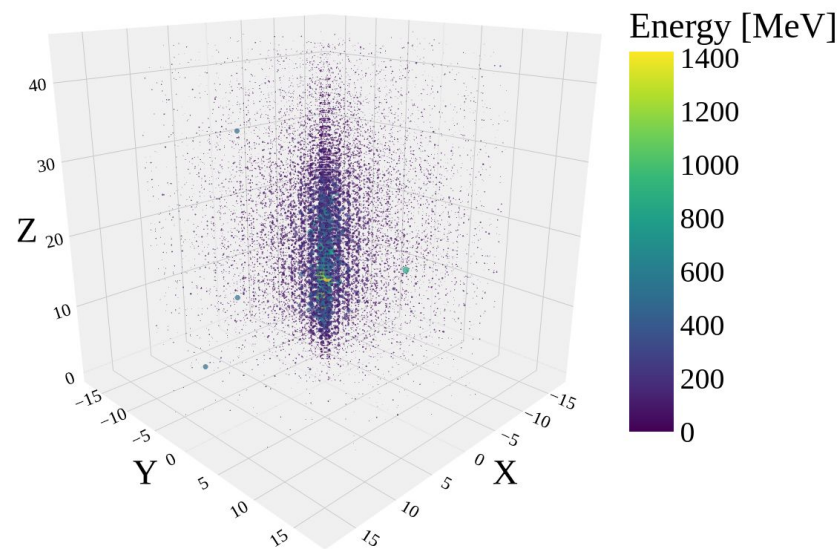
CaloChallenge

Results, Single Event Comparison

Geant 4



CaloCloud, time stamp: t

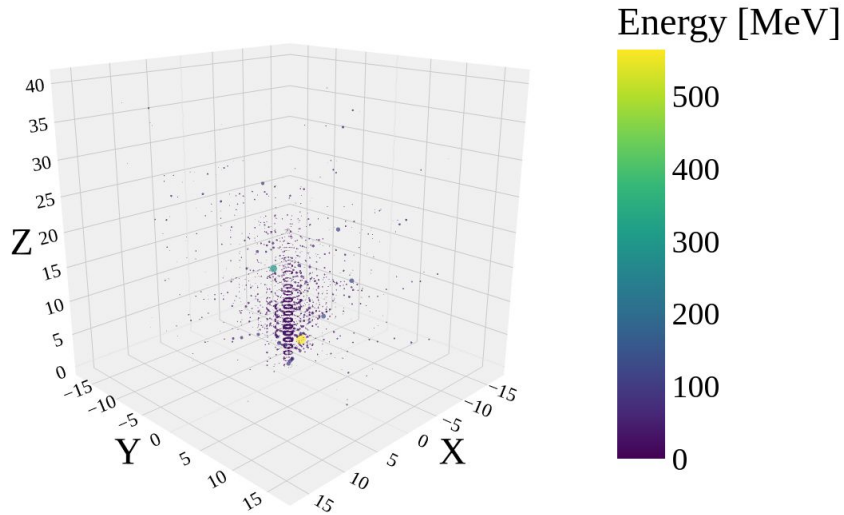


E = 1 TeV

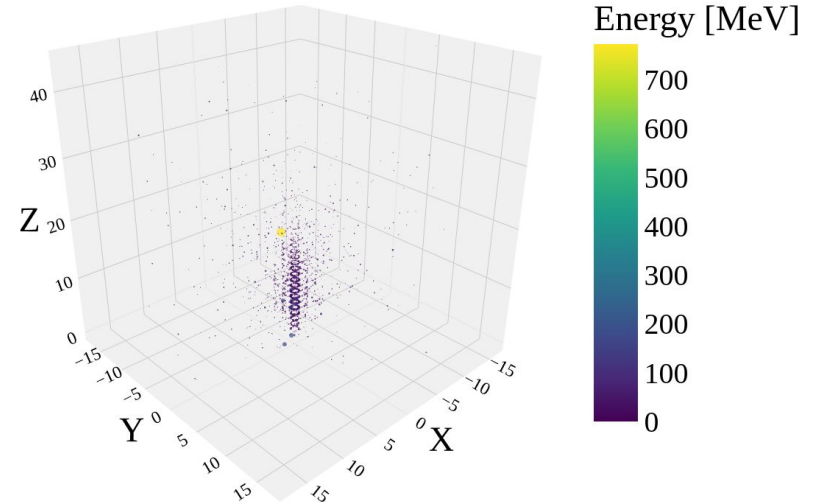
CaloChallenge

Results, Single Event Comparison

Geant 4



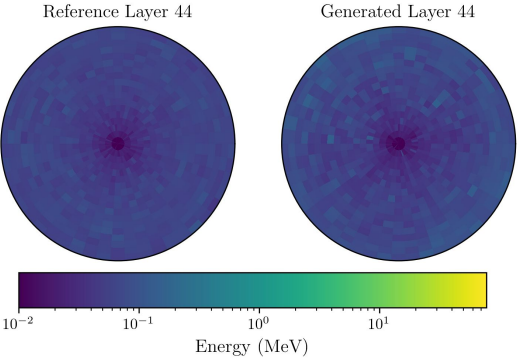
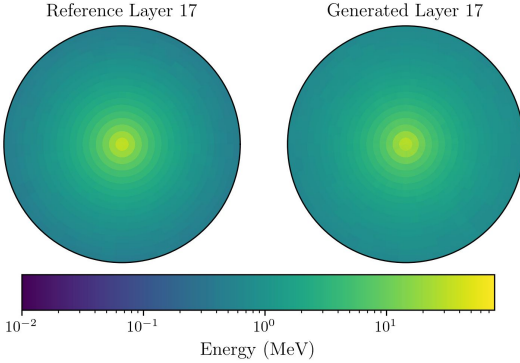
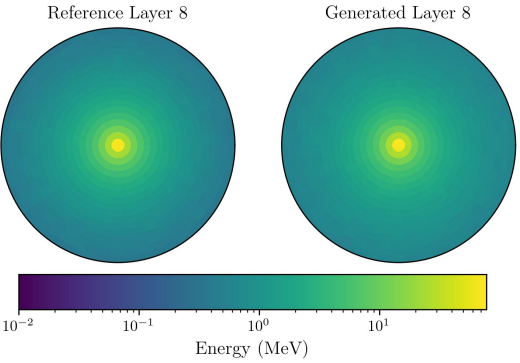
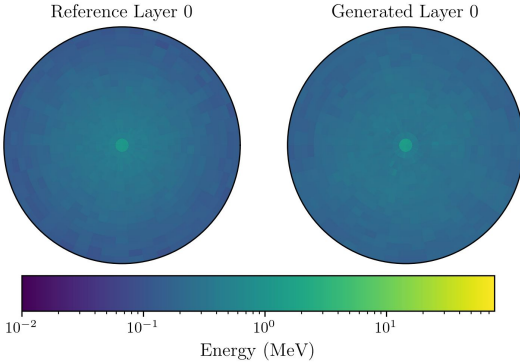
CaloCloud, time stamp: t_0



$E = 30 \text{ GeV}$

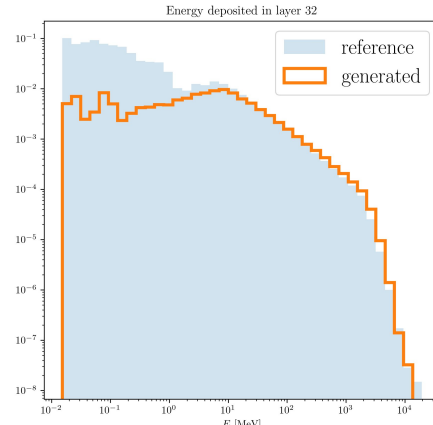
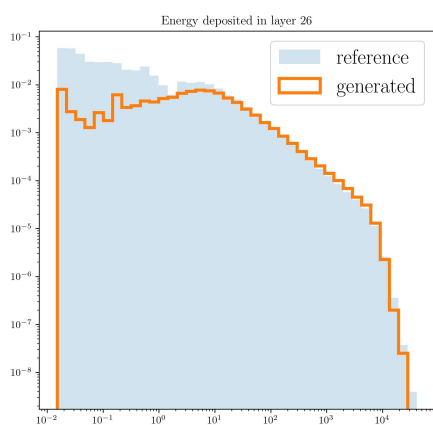
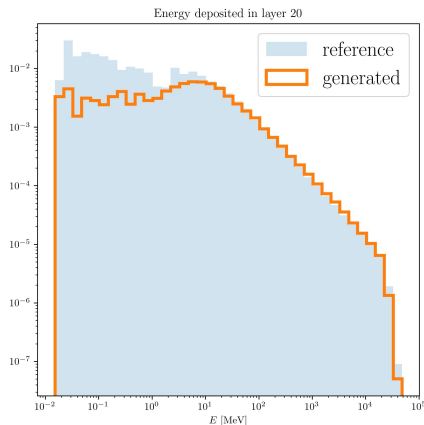
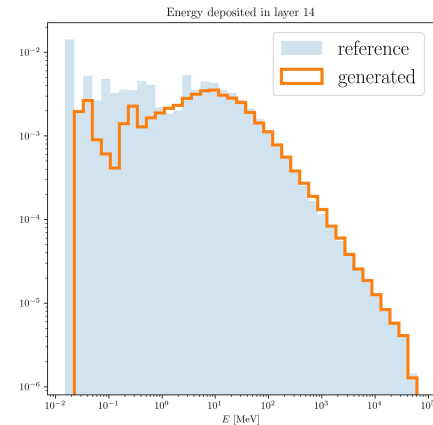
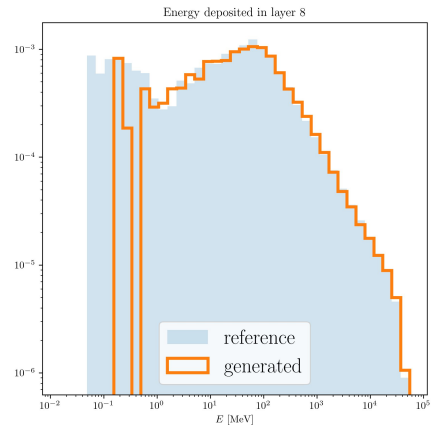
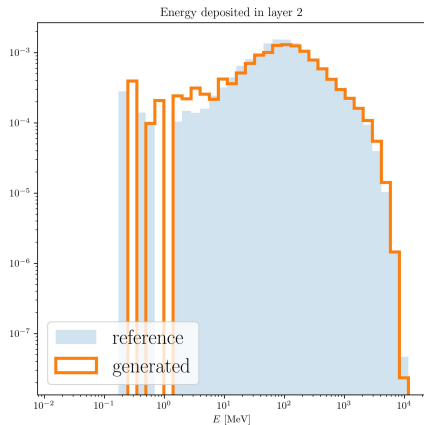
CaloChallenge

Results



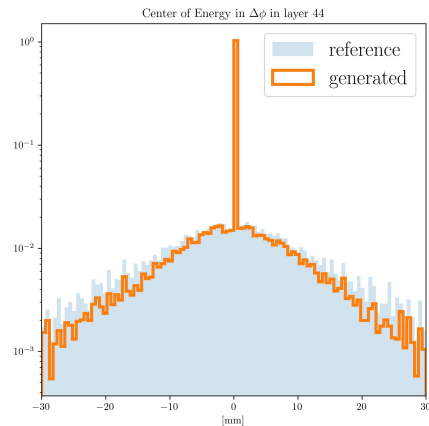
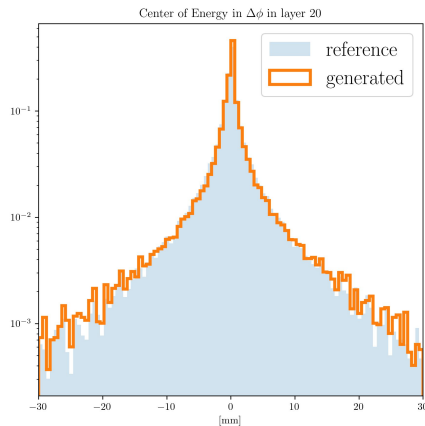
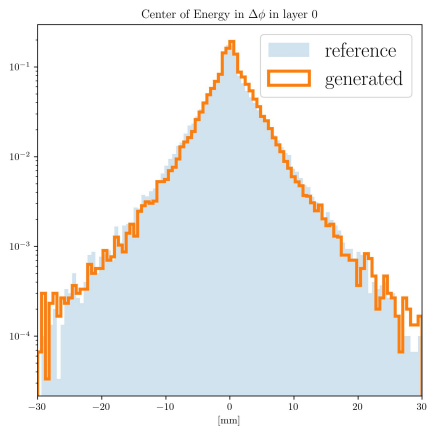
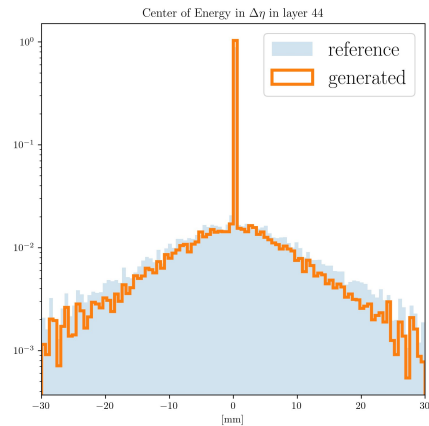
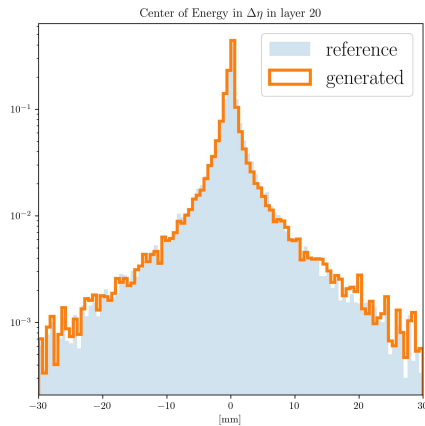
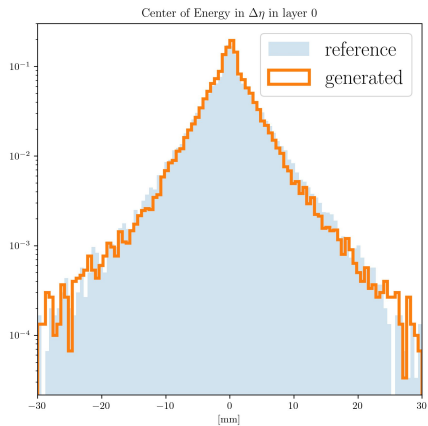
CaloChallenge

Results



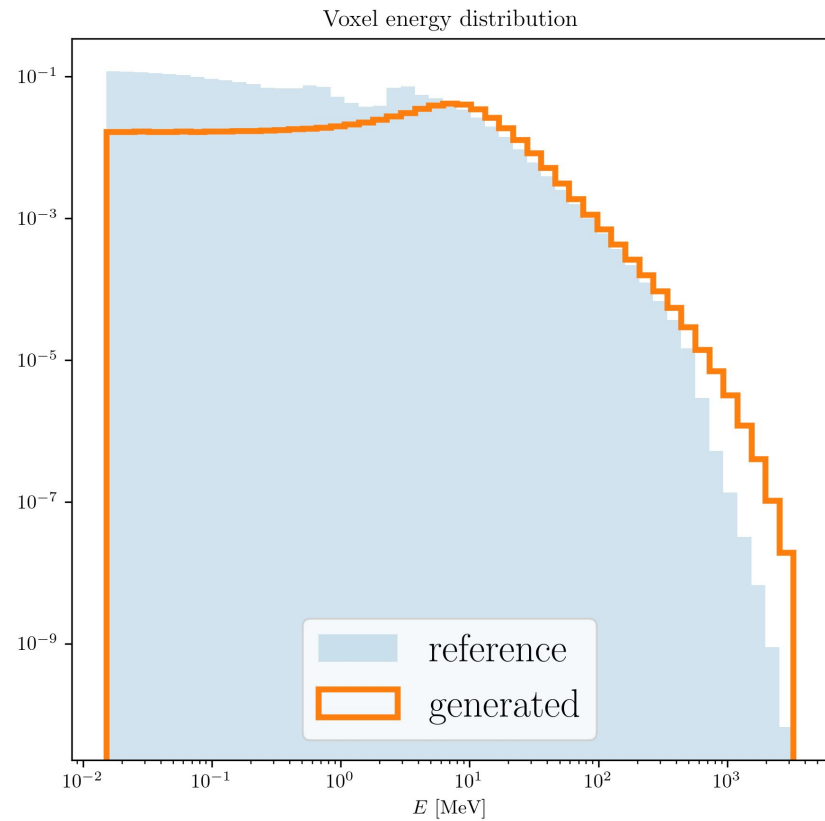
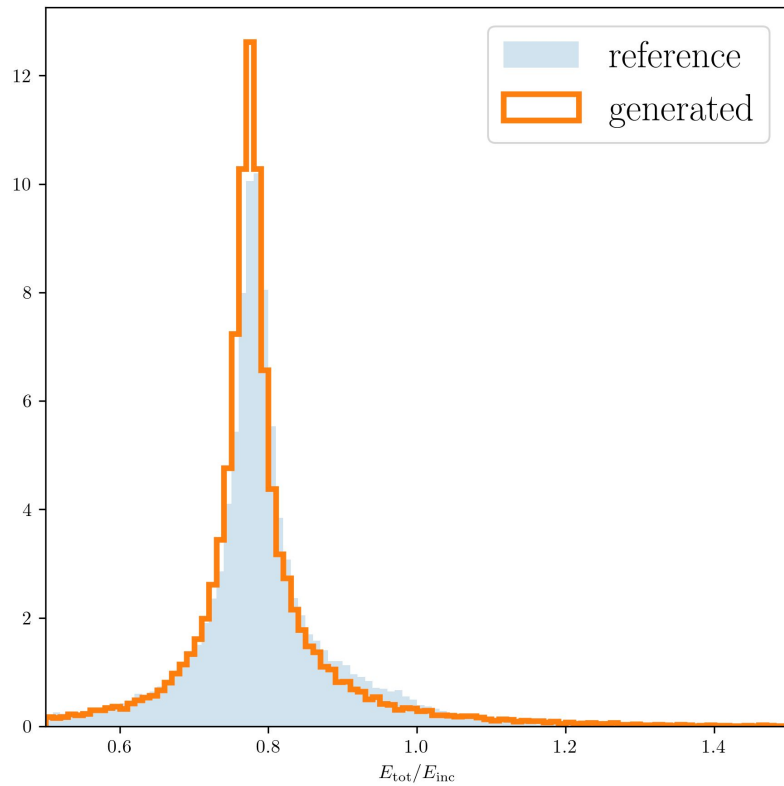
CaloChallenge

Results



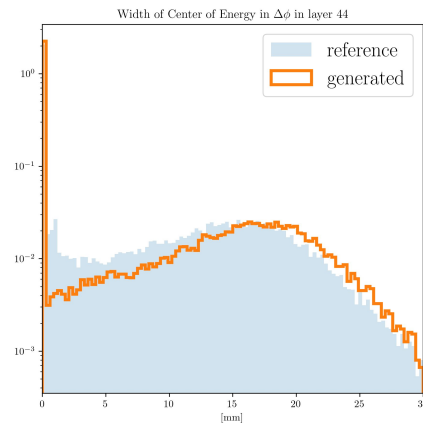
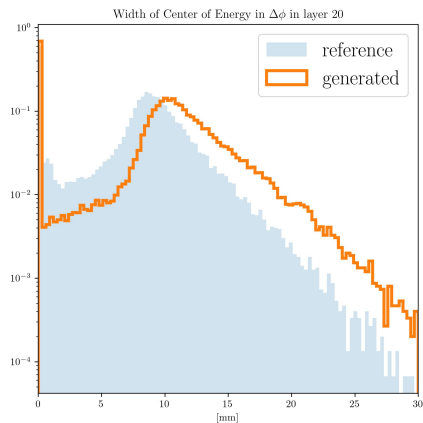
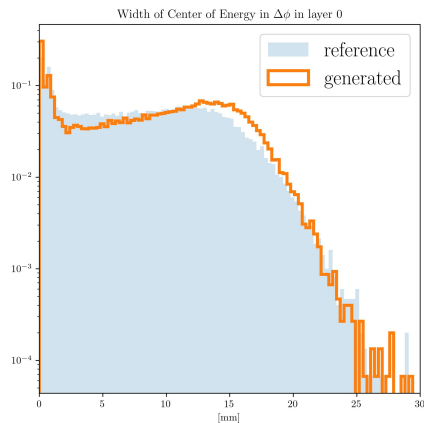
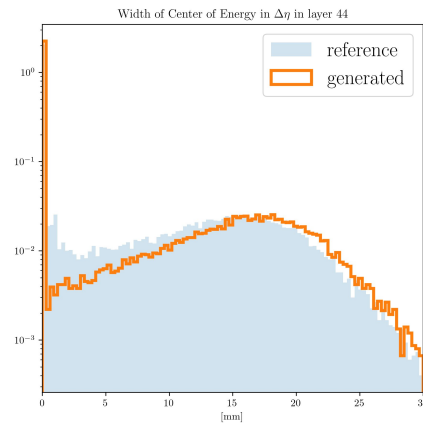
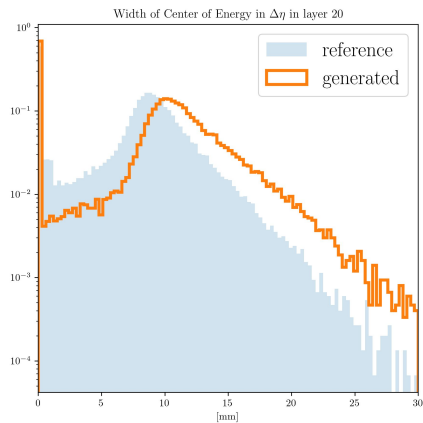
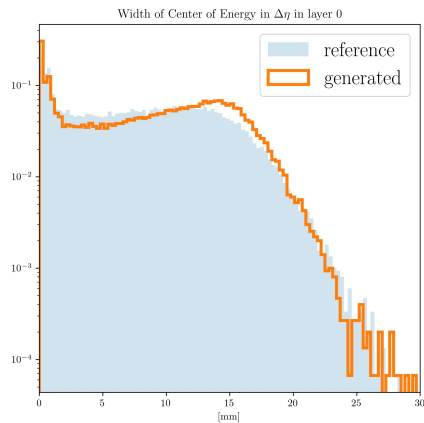
CaloChallenge

Results



CaloChallenge

Results



Summary

- Investigated new generative model architecture for generating EM showers data
- Fidelity of the physical properties learned well, but still have to be improved to achieve better agreement with GEANT4
- The combination of **Point Clouds** representation of EM showers and **Diffusion Model** looks promising as **a setup for easy integration into the simulation pipeline**
- Next step: work on timing performance, e.g. via distillation

BACKUP SLIDES

Point Cloud + Diffusion Model

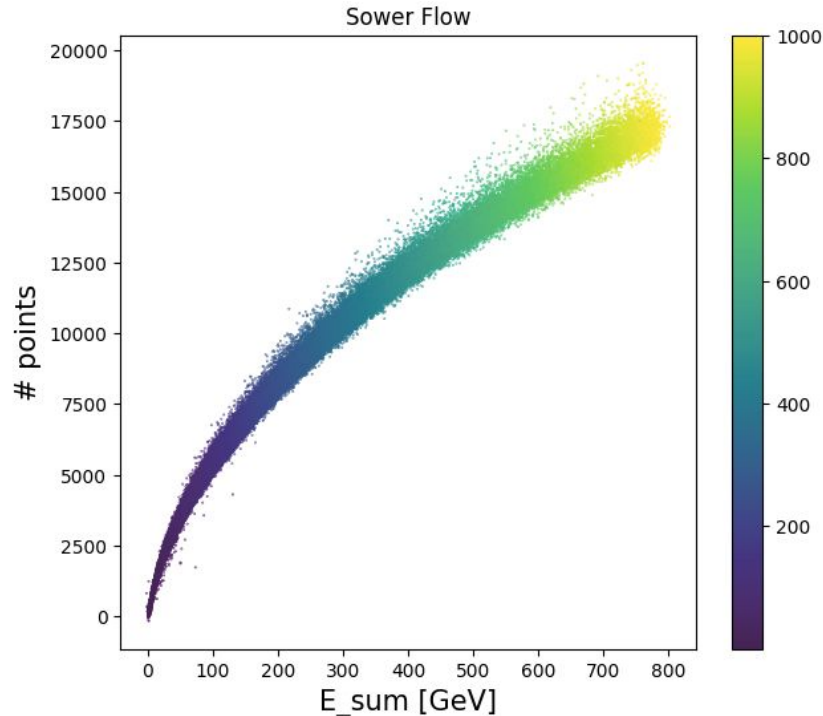
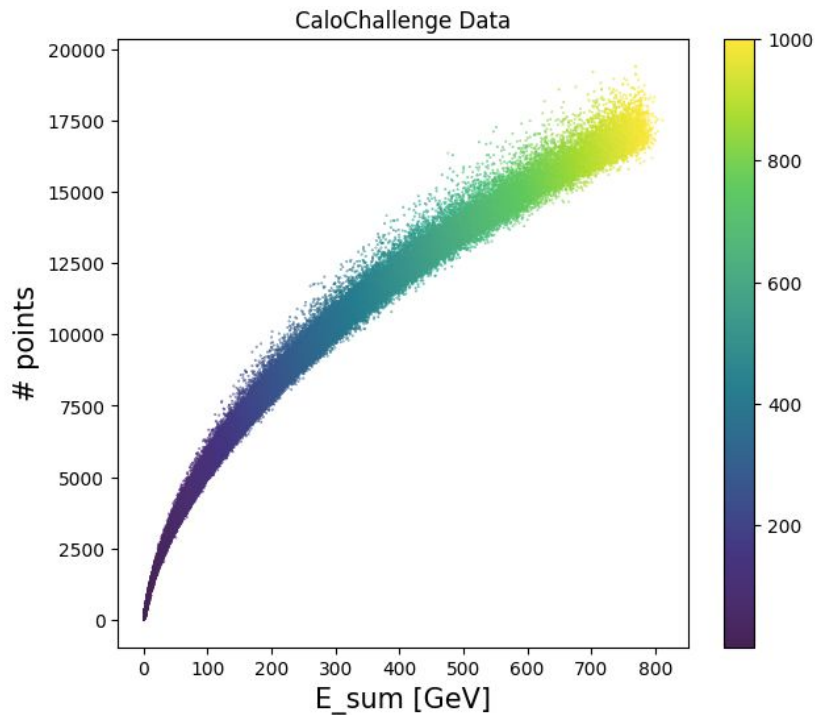
Results, Potential Speed-up

Hardware	Simulator	Time / Shower [ms]	Speed-up
CPU	GEANT4	4082 ± 170	$\times 1$
	CALOCLOUDS	3509 ± 220	$\times 1.2$
GPU	CALOCLOUDS	38 ± 3	$\times 107$

Timing performance needs to be improved in further work

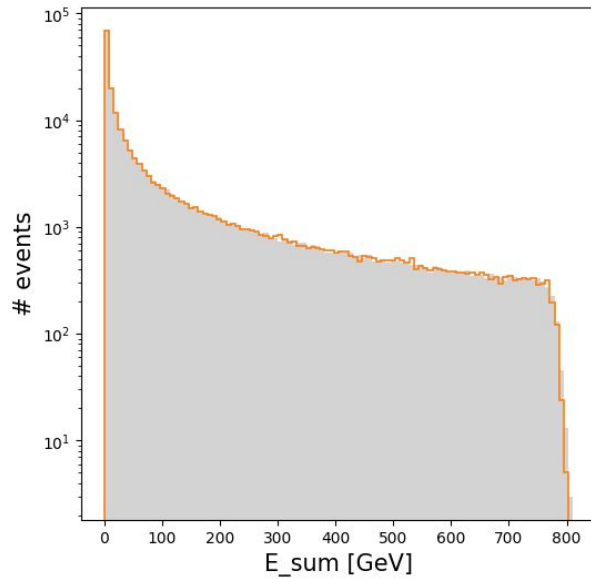
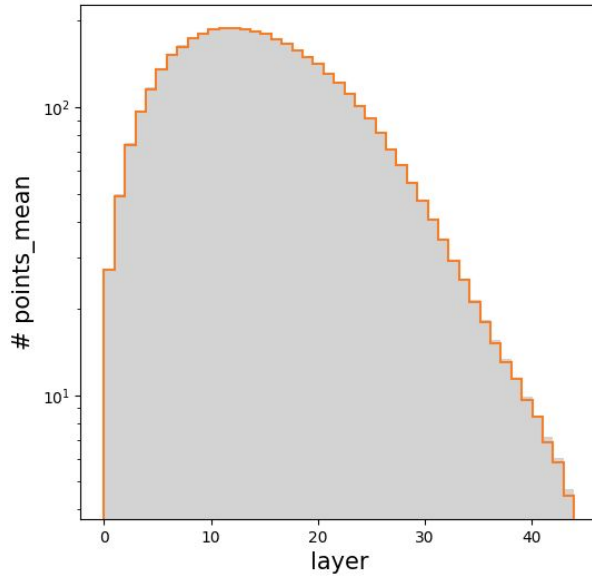
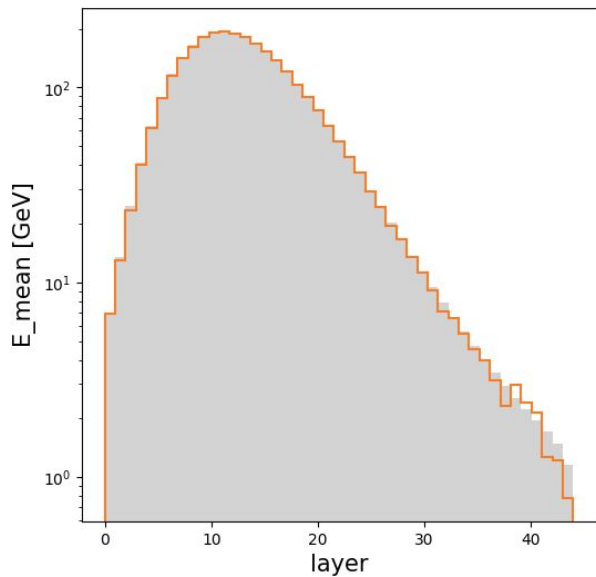
Shower Flow

Results



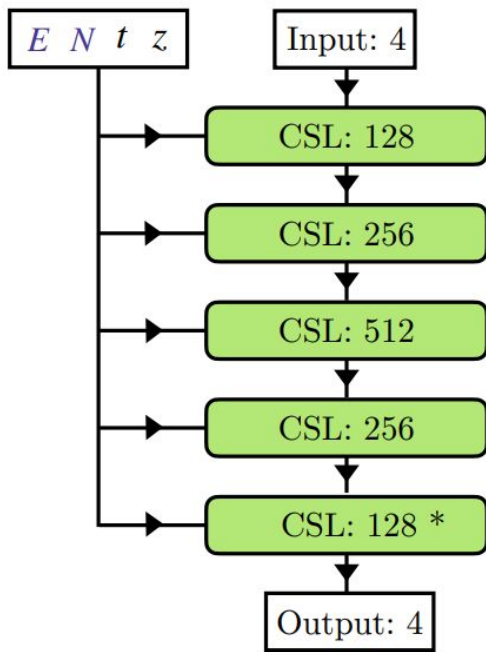
Shower Flow

Results

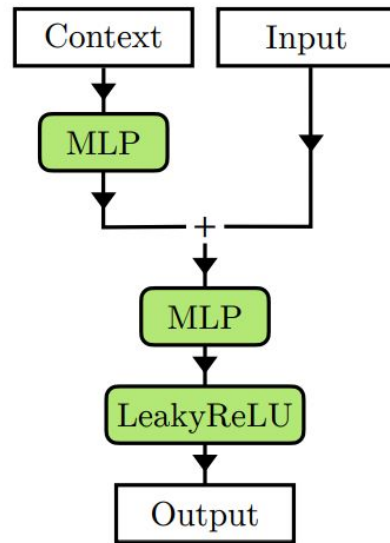


Point Cloud + Diffusion Model

PointWise Net



(a) PointWise Net

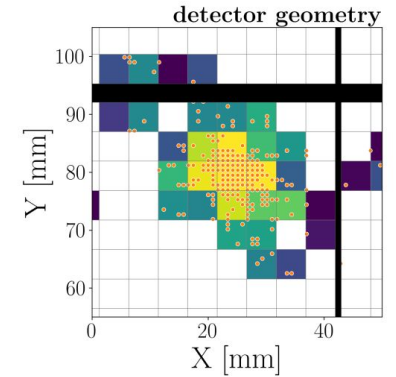
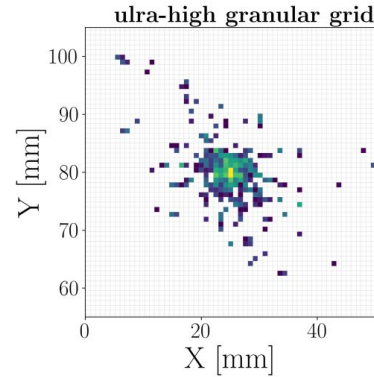
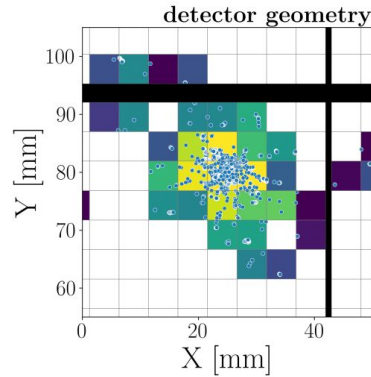


(b) ConcatSquash Layer (CSL)

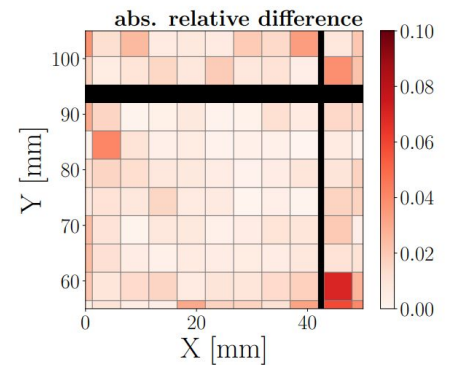
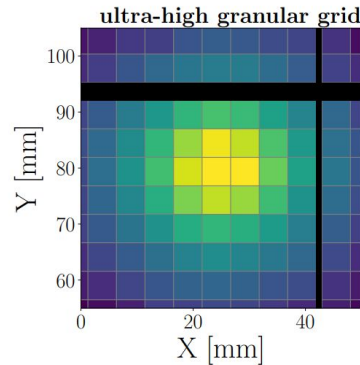
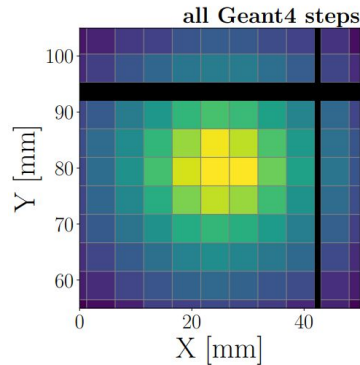
Point Cloud Representation of the EM Showers

Effects of the Pre-Clustering

Single event of 90 GeV shower in 21th layer



2k events of 90 GeV showers in 21th layer, overlay



Point Cloud Representation of the EM Showers

Effects of the Pre-Clustering

