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AllShowers: One model for all calorimeter showers

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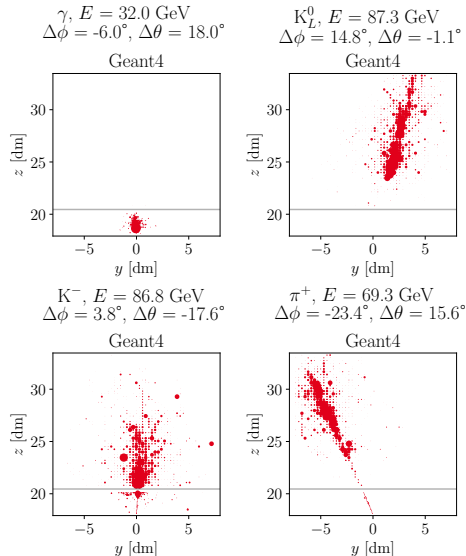
Motivation

Goal:

- ▶ previous models for specific particle types
 - ▶ CaloClouds¹: photons
 - ▶ CaloHadronic²: positively charged pions
- ▶ one model for all particle types
 - ▶ reduce memory footprint and complexity
 - ▶ leverage shared knowledge between different particle showers

Dataset:

- ▶ 12 particle types
- ▶ simulated with Geant4 and DD4hep
- ▶ ECAL + HCAL of International Large Detector (ILD)

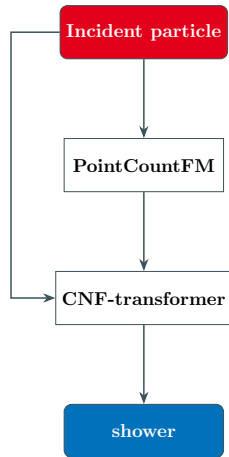


¹Erik Buhmann et al. CaloClouds: fast geometry-independent highly-granular calorimeter simulation. 2023. arXiv: 2305.04847

²Thorsten Buss et al. CaloHadronic: a diffusion model for the generation of hadronic showers. 2025. arXiv: 2506.21720

AllShowers Model Overview

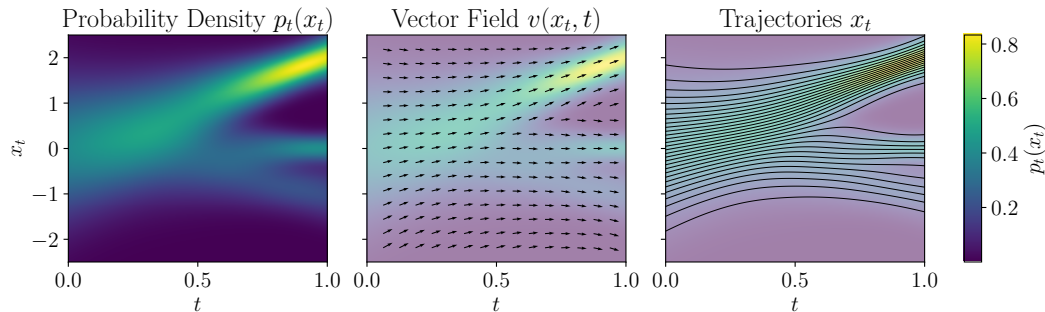
- ▶ continuous normalizing flow³
- ▶ based on point clouds
- ▶ attention mechanism for point interactions⁴
- ▶ conditioning on incident particle properties
 - ▶ particle type
 - ▶ energy
 - ▶ direction (normalized 3-momentum)
- ▶ PointCountFM generates the number of points per layer
- ▶ CNF-Transformer generates the point positions and energies
 - ▶ layer index given as a condition
 - ▶ one embedding vector per calorimeter layer
 - ▶ one embedding vector per particle type



³Yaron Lipman et al. Flow Matching for Generative Modeling. 2023. arXiv: 2210.02747

⁴Ashish Vaswani et al. Attention Is All You Need. 2017. arXiv: 1706.03762

Continuous Normalizing Flow



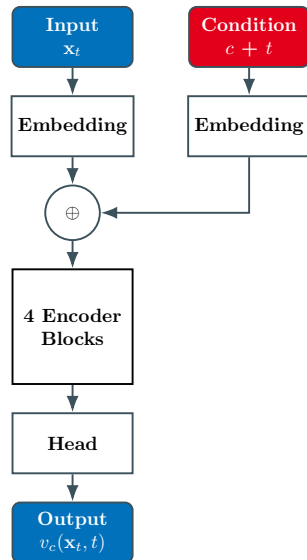
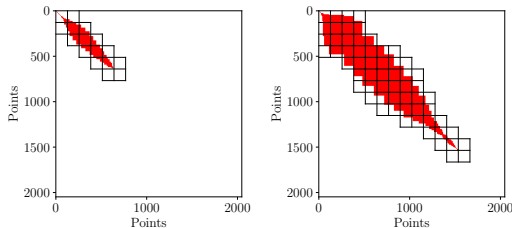
$$\partial_t \mathbf{x}_t = v(\mathbf{x}_t, t) \quad \mathbf{x}_0 \sim p_0$$

$$\log p_1(\mathbf{x}_1) = \log p_0(\mathbf{x}_0) - \int_0^1 \text{Tr} \left(\frac{\partial v}{\partial \mathbf{x}_t} \right) dt$$

- ▶ regressing $v(\mathbf{x}_t, t)$ with a network
 - ▶ training with flow matching
- ▶ sampling
 - ▶ sample noise from prior distribution
 - ▶ solve the ODE given by the network

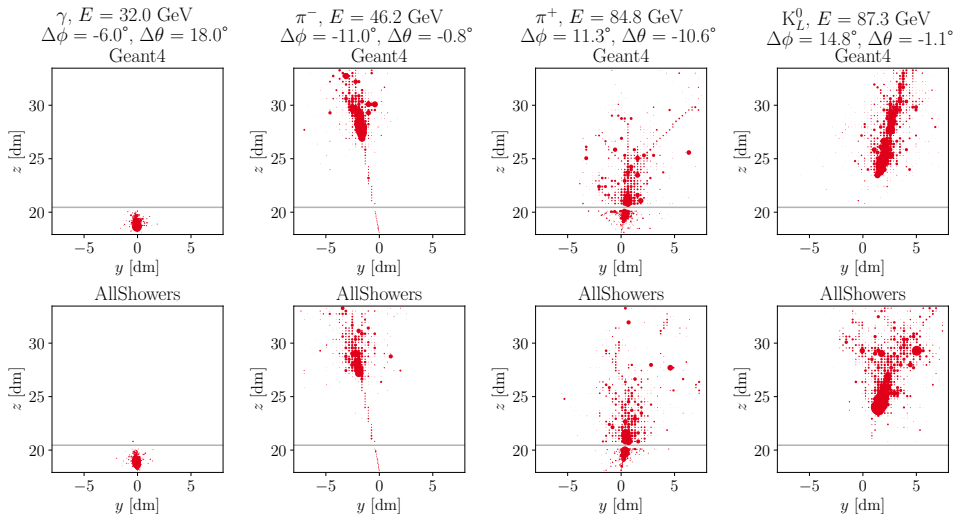
CNF-Transformer

- ▶ model learns $v_c(\mathbf{x}_t, t)$
- ▶ transformer encoder blocks with self-attention
- ▶ using flex-attention⁵
- ▶ only points up to two calorimeter layers apart can interact

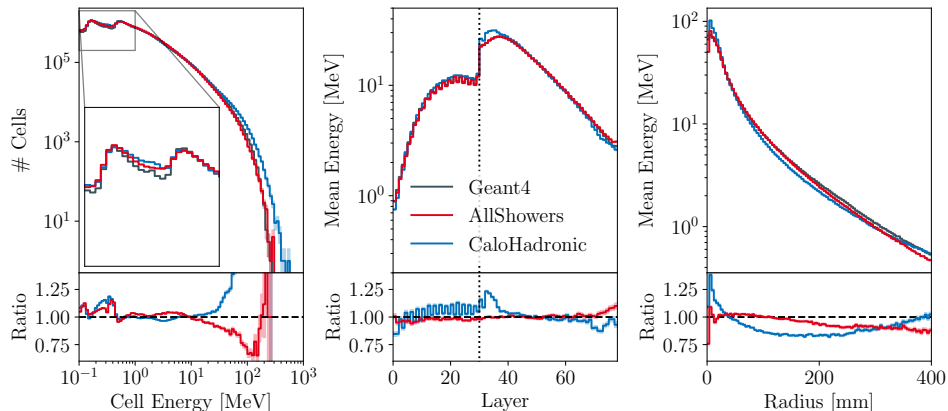


⁵ Juechu Dong et al. Flex Attention: A Programming Model for Generating Optimized Attention Kernels. 2024. arXiv: 2412.05496

Individual Showers

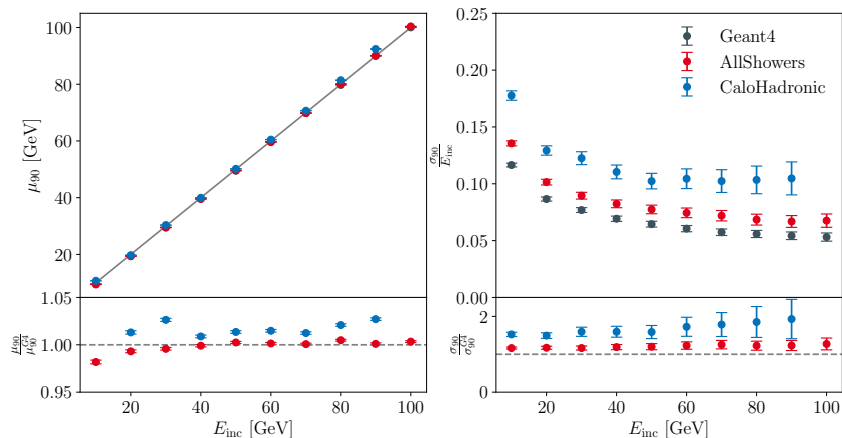


Comparison to CaloHadronic (Pions)



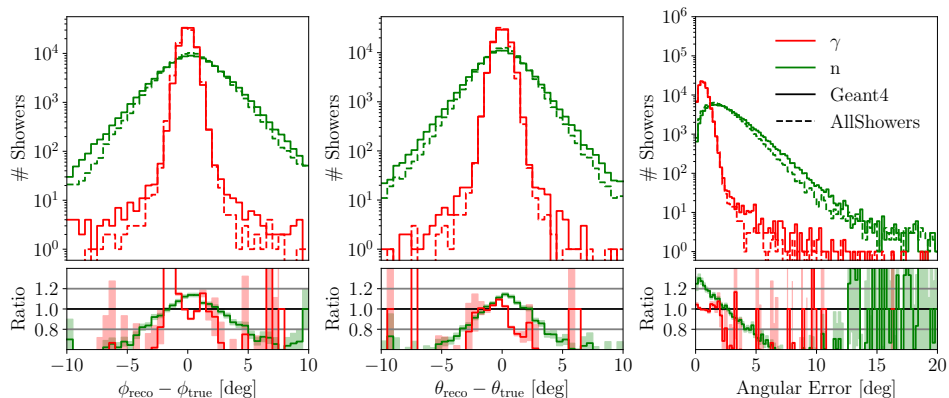
- hit energy, layer energy and radius distributions for π^+ showers
- comparison between AllShowers, CaloHadronic and Geant4

Linearity and Resolution (Pions)



- ▶ linearity and resolution for π^+ showers
- ▶ comparison between AllShowers, CaloHadronic and Geant4

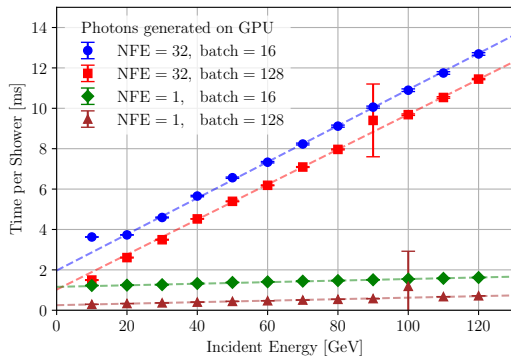
Internal Angle



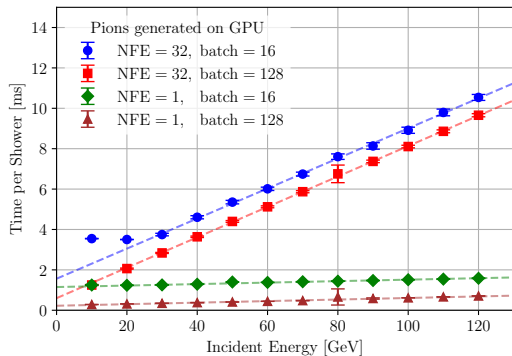
- PCA-based angular reconstruction
- overall good agreement with Geant4

- hadrons harder to reconstruct
- some larger deviations in the tails

Inference Time



- ▶ generation time on an NVIDIA A100
- ▶ linear scaling with incident energy



- ▶ slightly faster generation for pions
- ▶ constant offset independent of energy

Summary

- ▶ AllShowers is a generative model for calorimeter showers
- ▶ unifying generation of 12 particle types in one model
- ▶ conditioned on incident particle properties:
 - ▶ particle type
 - ▶ energy
 - ▶ direction (normalized 3-momentum)
- ▶ well reproduces sample-wise and distribution-wise features
- ▶ higher fidelity than shower-type-specific models for hadronic showers

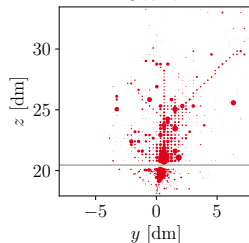


Paper on arXiv
2601.11716

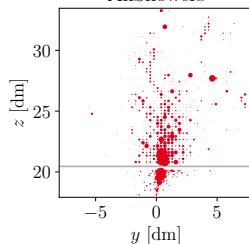


Code on GitHub
FLC-QU-hep/
AllShowers

π^+ , $E = 84.8$ GeV
 $\Delta\phi = 11.3^\circ$, $\Delta\theta = -10.6^\circ$
Geant4



AllShowers

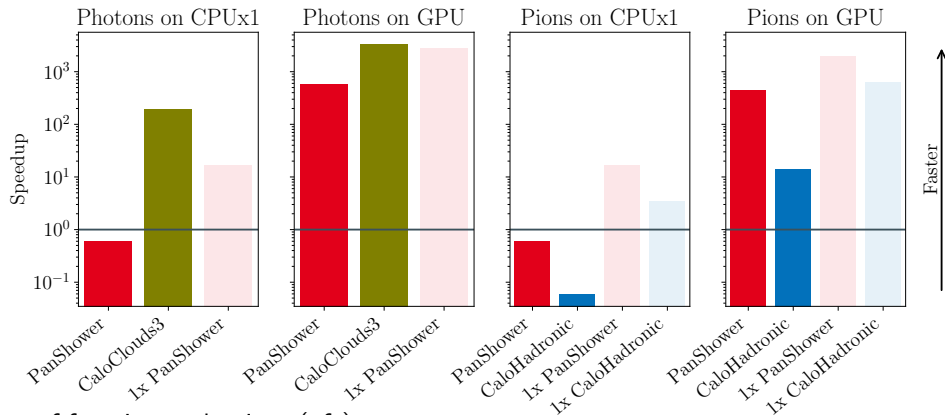


Model Size

Model	Layer-level model	Point-level model	Total
AllShowers	351,822	263,491	615,313
CaloClouds3	6,026,520	69,640	6,096,160
CaloHadronic	349,905	1,784,724	2,134,629

- ▶ Number of trainable parameters for AllShowers, CaloClouds3 and CaloHadronic
- ▶ AllShowers is significantly smaller than previous models

Inference Time

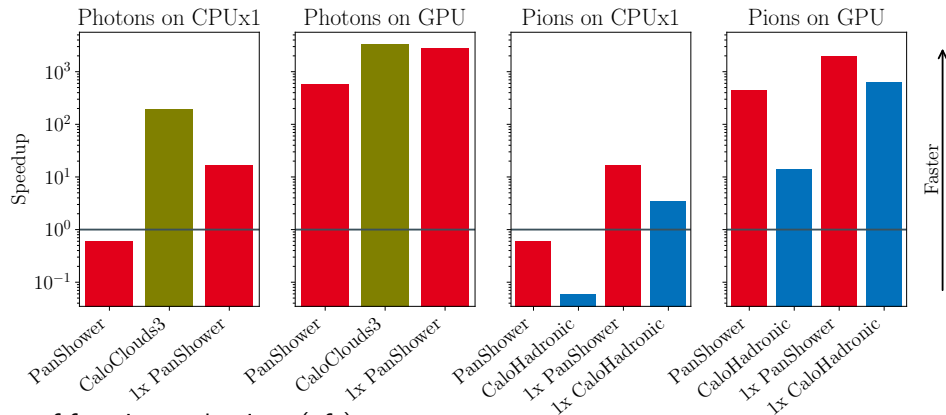


Number of function evaluations (nfe):

CaloClouds3	1		
AllShowers	32	1x AllShowers	1
CaloHadronic	59	1x CaloHadronic	1

	Hardware	batch size
CPU	AMD EPYC 7402	1
GPU	NVIDIA A100	16

Inference Time

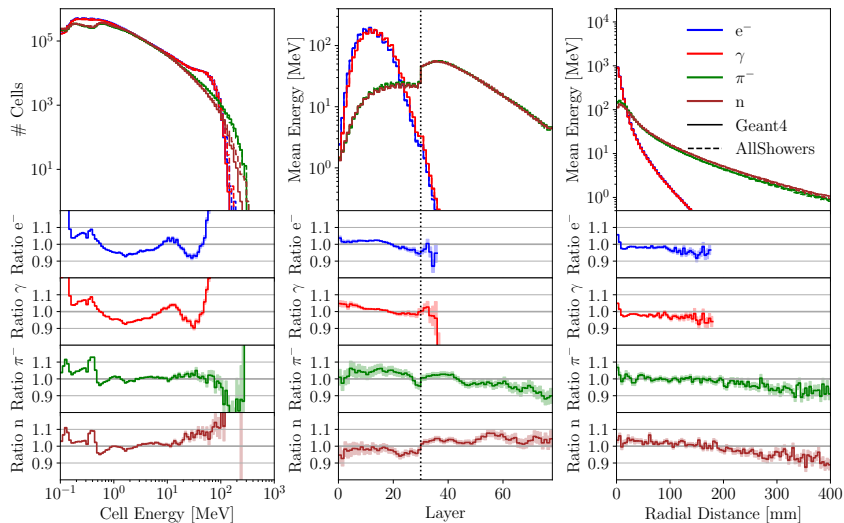


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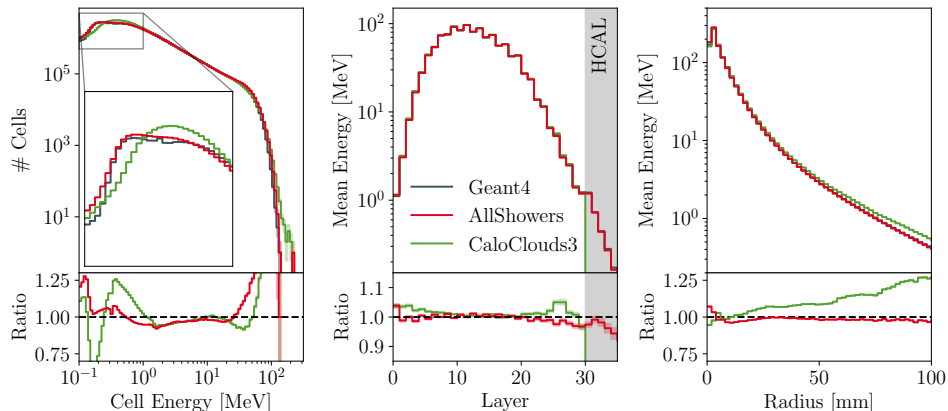
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Distribution Level Comparison

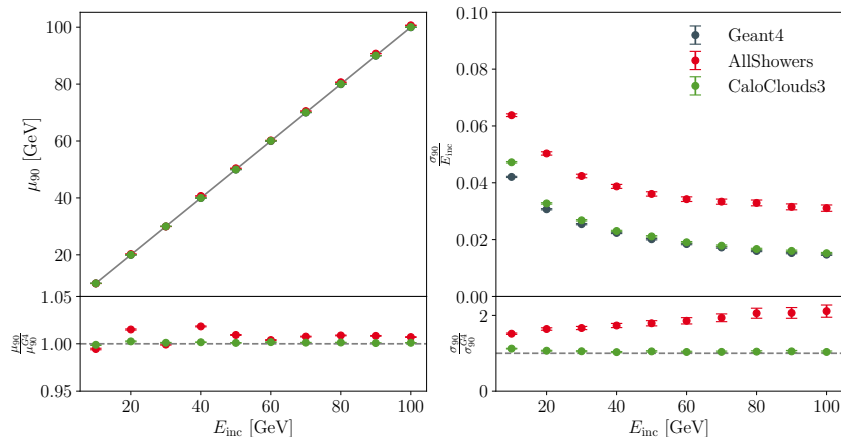


Comparison to CaloClouds3 (Photons)



- hit energy, layer energy and radius distributions for photon showers
- comparison between AllShowers, CaloClouds3 and Geant4

Linearity and Resolution (Photons)



- ▶ linearity and resolution for photon showers
- ▶ comparison between AllShowers, CaloClouds3 and Geant4