# **AIDAinnova 12.3 FastSim Activities at DESY**

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CLUSTER OF EXCELLENCE

QUANTUM UNIVERSE

HELMHOLTZ









# **Calo-ML Working Group at DESY**

- goal: study the suitability of using generative networks for the simulation of calorimeter showers in highly granular calorimeters with high fidelity
  - ideally such that they can replace G4 in full simulation for physics analyses for Future Higgs Factories
  - work carried out in context of the ILD detector concept for the ILC and with the CALICE collaboration
- generative ML working group:
  - part of FTX-SFT at DESY
    - 1 post doc, 2 PhD students
- joined with UHH (G.Kasieczka) in Quantum Universe cluster of excellence
- AIDAinnova, ACCLAIM (Helmholtz Innovation Pool), CDCS,...



### **Energy and Angular Conditioning**

- Photons incident at fixed position
- Extend BIB-AE architecture
- Vary incident energy and polar angle
  - Large training sample 500k showers
  - Uniform in [ 10-100 GeV, 30-90 deg ]
  - Test/validation samples at dedicated energies and angles







New Angles on Fast Calorimeter Shower Simulation, Diefenbacher, et al., 2023 MLST in press DOI 10.1088/2632-2153/acefa9, arXiv: 2303.18150

#### **Energy and Angular Conditioning, Performance After Reconstruction**







**Adding Another Angle** 

- Need to condition on energy, theta and phi for full application
- Extending phase space can be challenging



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**Two-Angle Conditioned BIB-AE** 

models: DDFastShowerML

· Development ongoing

DD4hep

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https://gitlab.desy.de/ilcsoft/ddfastshowerml

 Use fast sim hooks in DDG4/Geant4 Use realistic, detailed detector models · Currently only supports CPU

- used BIB-AE for fast simulation of photons from tau-pairs in ILD w/ DDML
- run full reconstruction (w/ PandoraPFA) of these events and compare G4 and BIB-AE

#### 20 GeV photon in ILD generated with a BIB-AE Integration into the Full Simulation Chain Prototype library for running ML-based fast sim Necessary update to Geant4 version 11.1! Trigger Model Fast Sim trigger Model-specific implementation of ML e.g. particle type, energy architecture · Aim to have an easy to use library which can be geometry adapted for all types of ML architectures in Inference Geometry Essential step to be able to study performance · Concrete inference in C++ Concrete placement in detector ONNX, LibTorch etc... · Endcap, barrel etc.

#### **P.McKeown**



of model with full physics benchmarks



- reasonably good description of pi0s from tau decays
- differences between BIB-AE of similar order as G4-10.4 (used for training) and G4-11.1 (used for reco)

# **Point Cloud Representation of the EM Showers**

Jy [MeV]

#### **GEANT4 Steps**

Photon Energy: 90 [GeV] Event: 4

Time step: 0.98246 [ns]

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

- All G4 interactions, ultimate resolution
- Detached from detector layer geometry







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CaloClouds: Fast Geometry-Independent Highly-Granular Calorimeter Simulation, Buhmann, et al. 2023, <u>arXiv:2305.04847</u>

**CaloClouds, Model Overview** 



(a) Training at random time step t



(b) Sampling with reverse diffusion through all time steps 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** Results

CaloClouds: Fast Geometry-Independent Highly-Granular Calorimeter Simulation, Buhmann, et al. 2023, <u>arXiv:2305.04847</u>



CaloClouds II: Ultra-Fast Geometry-Independent Highly-Granular Calorimeter Simulation, Buhmann, et al. 2023, <u>arXiv:2309.05704</u>

**CaloClouds II, Model Overview** 



Modified version of CaloClouds + Consistency Distillation  $\rightarrow$  significantly reduced inference time

## **Point Cloud + Diffusion Model** Results

CaloClouds II: Ultra-Fast Geometry-Independent Highly-Granular Calorimeter Simulation, Buhmann, et al. 2023, <u>arXiv:2309.05704</u>



## **Point Cloud + Diffusion Model** Results

**CaloClouds II: Ultra-Fast Geometry-Independent Highly-Granular Calorimeter Simulation**, Buhmann, et al. 2023, <u>arXiv:2309.05704</u>

Hardware	Simulator	NFE	Batch Size	Time / Shower [ms]	Speed-up
CPU	Geant4		3	$3914.80 \pm 74.09$	×1
	CALOCLOUDS	100	1	$3146.71 \pm 31.66$	$\times 1.2$
	CALOCLOUDS II	25	1	$651.68 \pm 4.21$	$\times 6.0$
	CaloClouds II (CM)	1	1	$84.35 \pm 0.22$	×46
GPU	CALOCLOUDS	100	64	$24.91 \pm 0.72$	$\times 157$
	CALOCLOUDS II	25	64	$6.12 \pm 0.13$	$\times 640$
	CaloClouds II (CM)	1	64	$2.09\pm0.13$	$\times 1873$

Work in Progress, Adding Two-Angle Conditioning



 $50 \,\, {\rm GeV}$ 

50 GeV

# Summary

**BiB-AE** Family

CaloClouds Family

Flows Family

Getting High: High Fidelity Simulation of High Granularity Calorimeters with High Speed, Buhmann et al., <u>arXiv:2005.05334</u>, Comput Softw Big Sci 5, 13 (2021) New Angles on Fast Calorimeter Shower Simulation, Diefenbacher, et al., 2023 MLST in press DOI 10.1088/2632-2153/acefa9, arXiv: 2303.18150

**Hadrons, Better, Faster, Stronger** Buhmann, et al., **arXiv:2112.09709**, MLST 3 2, 025014 (2022) High Fidelity / High speed / Challenging to Scale / Challenging to Integrate

CaloClouds: Fast Geometry-Independent Highly-Granular Calorimeter Simulation, Buhmann, et al. 2023, <u>arXiv:2305.04847</u> **CaloClouds II: Ultra-Fast Geometry-Independent Highly-Granular Calorimeter Simulation**, Buhmann, et al. 2023, <u>arXiv:2309.05704</u>

# **Fair Fidelity** / High speed / Easy to Scale / Layer Geometry Independent / Straightforward to Integrate

Convolutional L2LFlows: Generating Accurate Showers in Highly Granular Calorimeters Using Convolutional Normalizing Flows, Buss et al., coming soon on arXiv...

L2LFlows: Generating High-Fidelity 3D Calorimeter Images, Diefenbacher et al., <u>arXiv:2302.11594</u>

#### Ultimate Fidelity / Fair speed / Challenging to Scale / Layer Geometry Independent / Straightforward to Integrate

# **BACKUP SLIDES**

# Image Representation of the EM Showers



## Problems with Image Representation of the EM Showers ILD Detector, ECAL Layers Structure



## **Problems with 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 Data Preprocessing



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

## New Angles Energy Conditioning Performance

• Sim level visible energy

- **Rec** level calibrated energy
  - After full PandoraPFA reco



#### New Angles on Fast Calorimeter Shower Simulation, Diefenbacher, et al. 2023 MLST in press DOI 10.1088/2632-2153/acefa9, arXiv: 2303.18150



## **New Angles Angular Conditioning Performance**

0.35

0.30

0.25

0.15

0.10

0.05

0.00

0.35

0.30

0.25

0.15

0.10

0.05

0.00

normalized 0.20

normalized 0.20 40 degree 20 GeV Photons

- Geant4

-- BIB-AE PP

Sim Level

60 degree

- 85 degree

60 degree

85 degree

Sim level angle reconstruction

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- **Rec** level angle reconstruction
  - After full reconstruction with PandoraPFA



0.5

0.4

0.3

0.2

normalized

60 degree

85 degree

#### New Angles on Fast Calorimeter Shower Simulation, Diefenbacher, et al. 2023 MLST in press DOI 10.1088/2632-2153/acefa9, arXiv: 2303.18150



## Shower Flow Results



## Shower Flow Results



**PointWise Net** 



**Results, Position of the Center of Gravity** 



**Results, Visible Energy and the Number of Hits** 



### Point Cloud + Diffusion Model Results

CaloClouds: Fast Geometry-Independent Highly-Granular Calorimeter Simulation, Buhmann, et al. 2023, <u>arXiv:2305.04847</u>



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

## Point Cloud Representation of the EM Showers Effects of the Pre-Clustering



# **Point Cloud Representation of the EM Showers**

**Effects of the Pre-Clustering** 



