

AIDAInnova - 12.3 FastSim Activities at DESY

Annual Meeting 2023

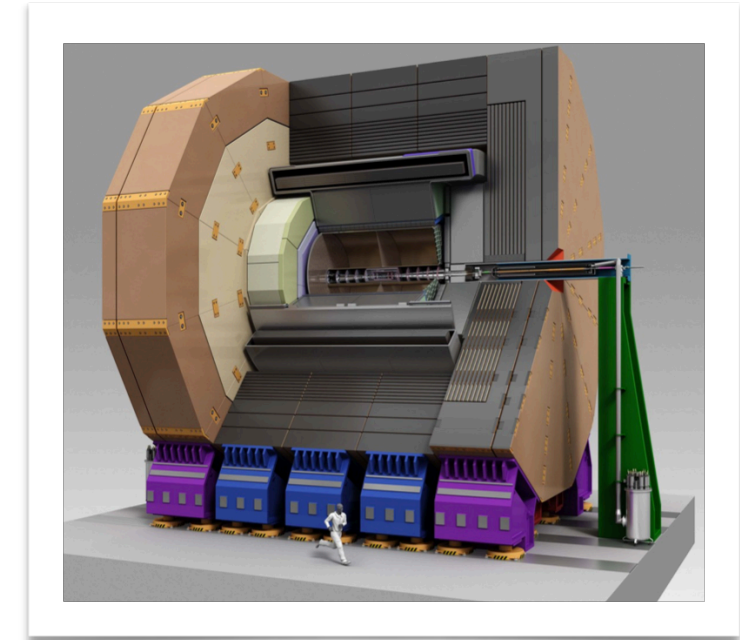
Valencia

F. Gaede, DESY
24.04.2023

Calo-ML Working Group at DESY



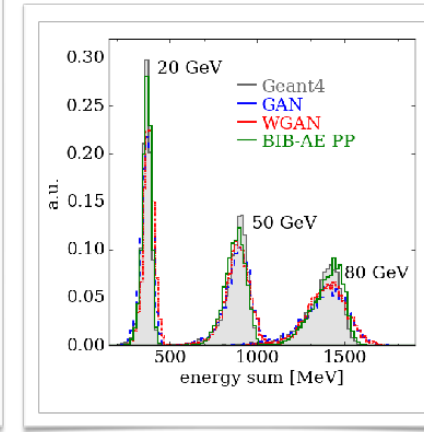
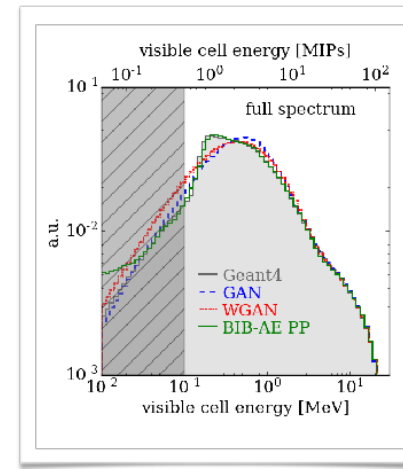
- goal: study the suitability of using **generative networks** for the simulation of **calorimeter showers** in **highly granular** calorimeters with **high fidelity**
 - ideally such that they **can replace G4 in full simulation for physics analyses** for Future Higgs Factories
 - work carried out in 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, 3 PhD students
 - joined with **UHH** (G.Kasieczka) in **Quantum Universe** cluster of excellence
 - **AIDAinnova**, ACCLAIM (Helmholtz Innovation Pool), CDCS,...



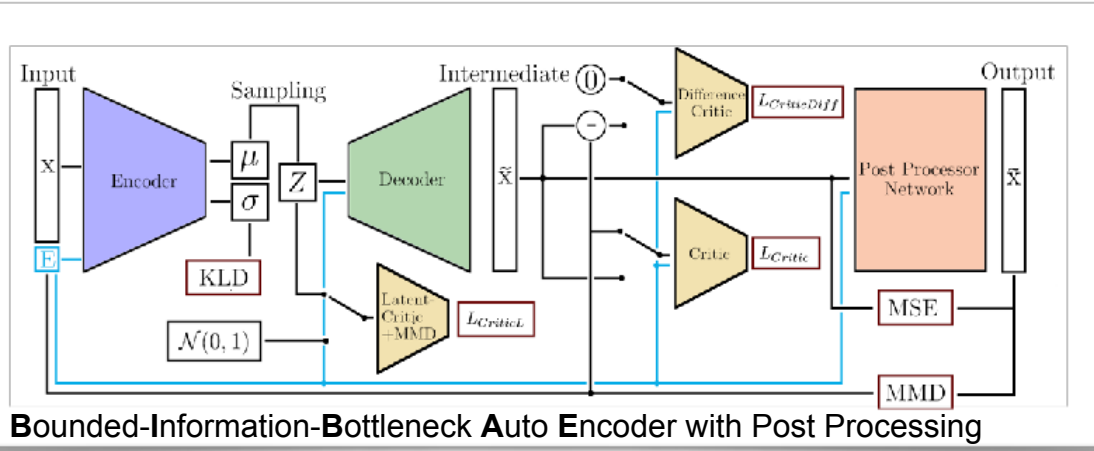
Previous Work

High fidelity simulation of Photons

- started with photons at **90 deg impact angle** in ILD Ecal (5x mm², SiW) w/ uniform energies 10-100 GeV
- achieved **high fidelity** in distributions of relevant physical variables
- using **Bounded-Information-Bottleneck Auto Encoder (BIB-AE)** w/ post-processing
- extend approach to hadrons, different angles, reco, ...



Getting High: High Fidelity Simulation of High Granularity Calorimeters with High Speed, E. Buhmann et al. e-print: [2005.05334](https://arxiv.org/abs/2005.05334), published in Comput Softw Big Sci 5, 13 (2021)



Bounded-Information-Bottleneck Auto Encoder with Post Processing

From Photons to Pions

	Energy	Angles θ, ϕ	ECAL +HCAL	Reco
Photon showers • Predominantly governed by EM interactions • Compact structure Relatively easy to generalise	✓	+	N/A	+
Pion showers • Hadronic and EM interactions • Complex structure • Large event-to-event fluctuations Hard to learn	+	x	x	+

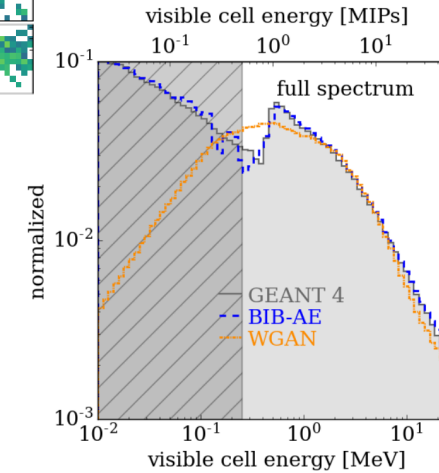
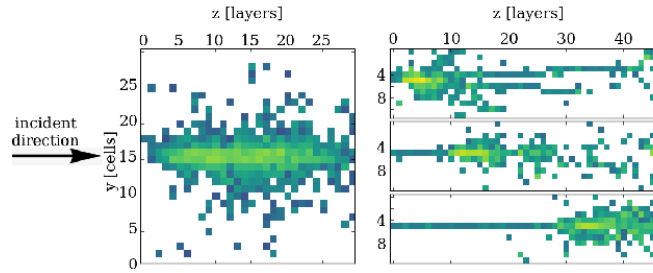
✓ = Achieved x = Yet to be done
 + = Addressed here

Previous Work

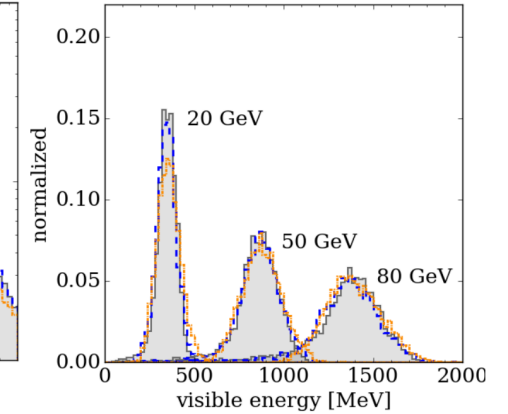
Hadronic Shower Simulation



- use charged pion sample in ILD SciFe AHcal
- 500k showers [10-100GeV] @ 90 deg
 - **no Ecal in front**
- projected into 25x25x48 grid
- investigate **WGAN** and **BIB-AE**
- achieve ~high fidelity for pion showers in highly granular calorimeter
- applied full PandoraPFA **reconstruction** to single showers
- eventually the important benchmark for ML based fast sim....



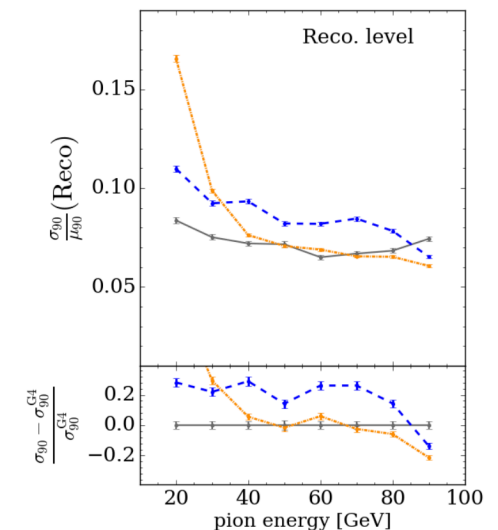
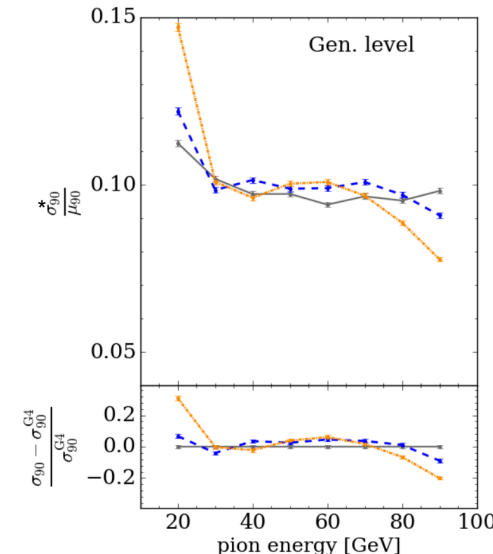
Very good agreement of MIP peak for **BIB-AE** with Post-Processing!



Great agreement with Geant4

Hadrons, Better, Faster, Stronger

E. Buhmann, S. Diefenbacher, E. Eren, F. Gaede et al,
 Published in: Mach.Learn.Sci.Tech. 3 (2022) 2, 025014,
[arXiv:2112.09709](https://arxiv.org/abs/2112.09709)

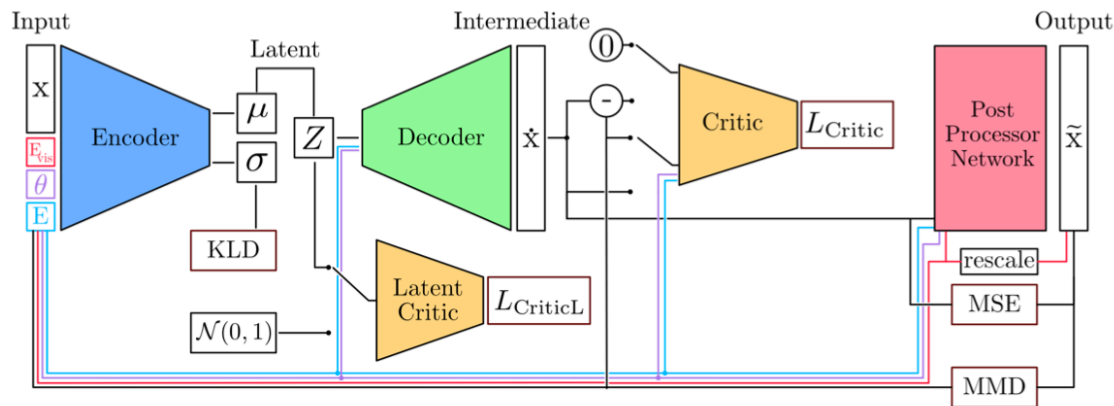
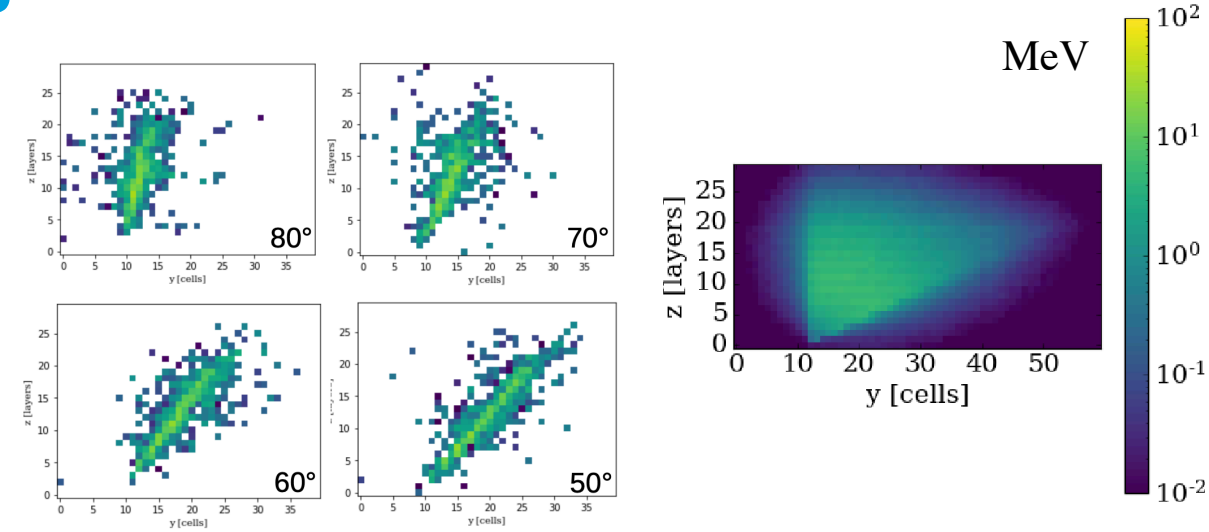


Energy and Angular Conditioning

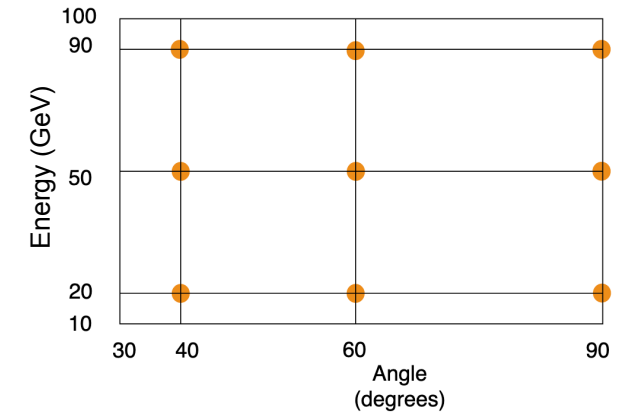
P.McKeown



- using photon sample in ILD SiW Ecal
- **angular and energy conditioning**
- large training sample, uniform in [10-100GeV, 30-90deg]
- train a **BIB-AE** w/140 k showers (500 k used for PP)
- significant GPU resources needed for training



New Angles on Fast Calorimeter Shower Simulation,
S.Diefenbacher, P.McKeown et al
[arXiv: 2303.18150](https://arxiv.org/abs/2303.18150), submitted to MLST

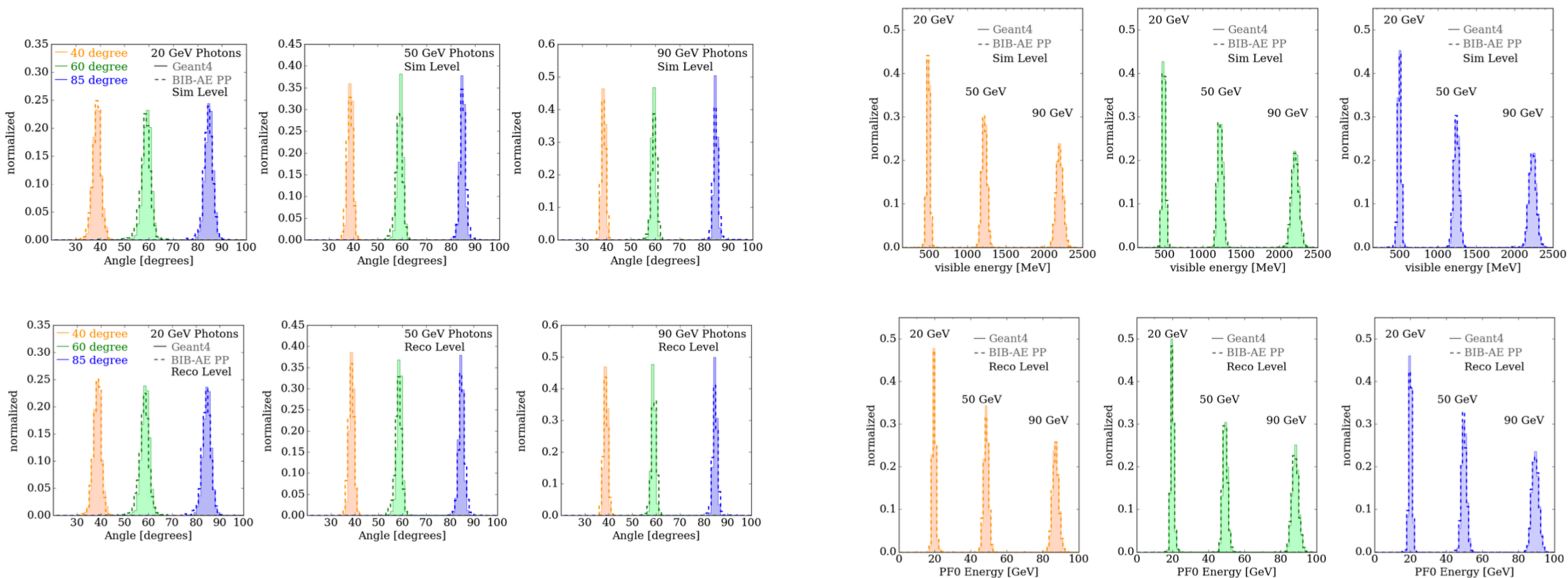


training and validation samples - 30x60x30 grid

Energy and Angular Conditioning

P.McKeown

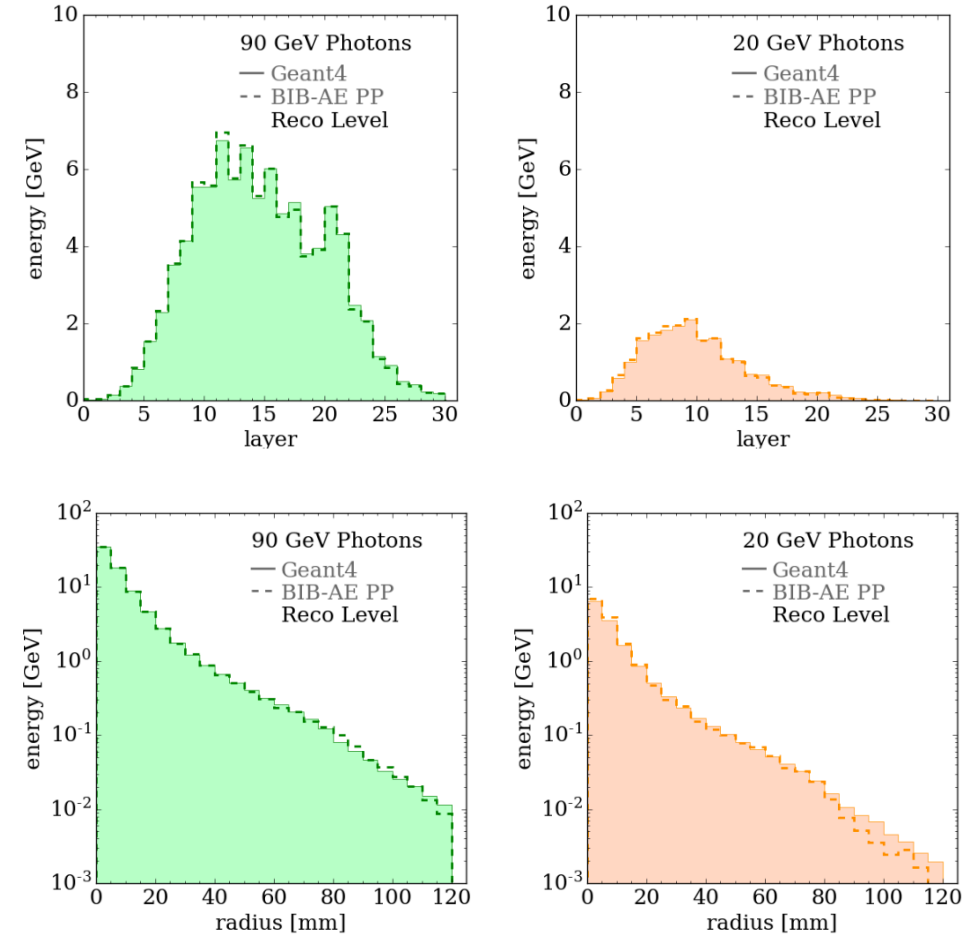
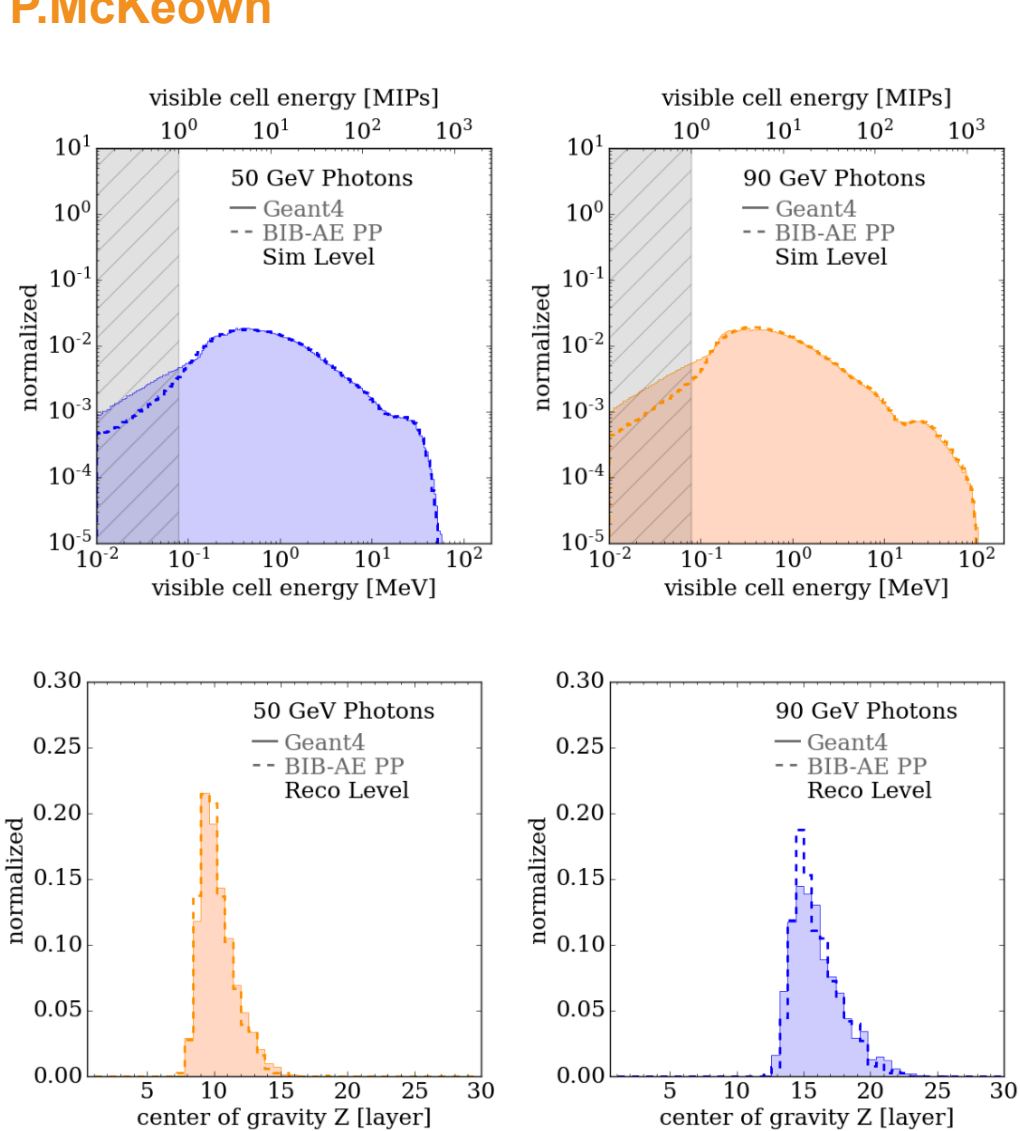
after simulation/generation



after full reconstruction w/ PandoraPFA

Energy and Angular Conditioning

P.McKeown

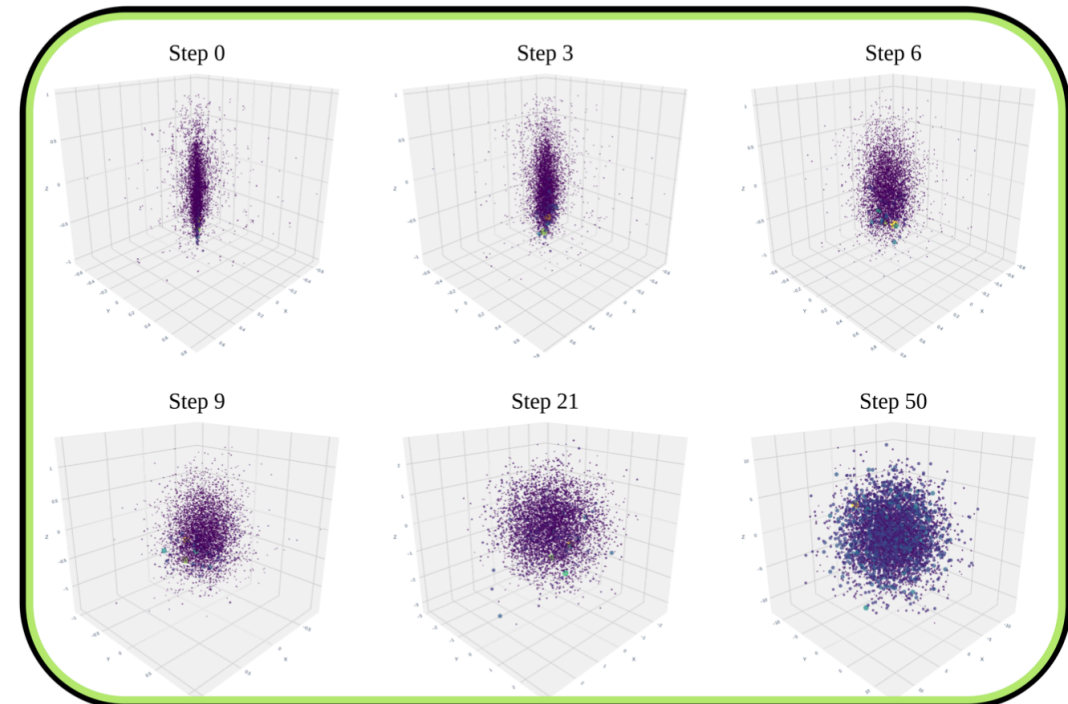
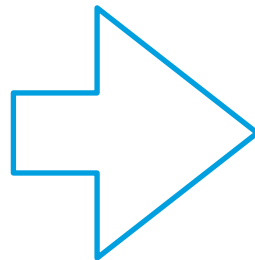
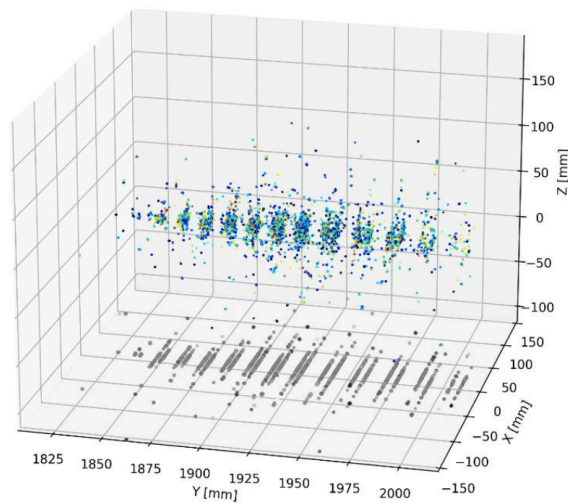
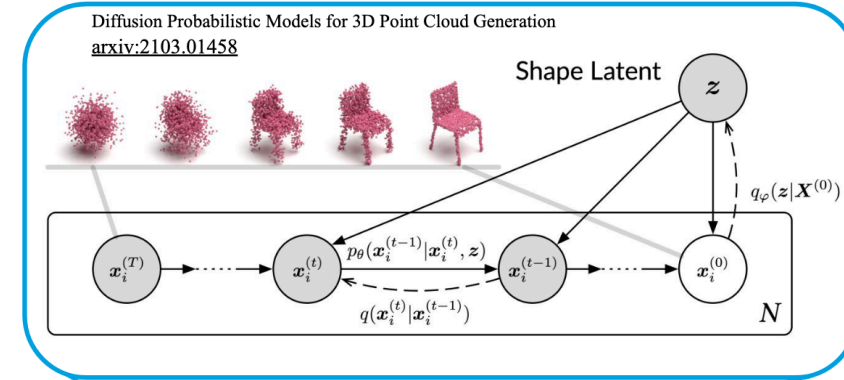


- some example distributions - best (left) and worst (right) in validation data set
- achieve excellent physics fidelity throughout

Point Cloud Diffusion

A.Korol

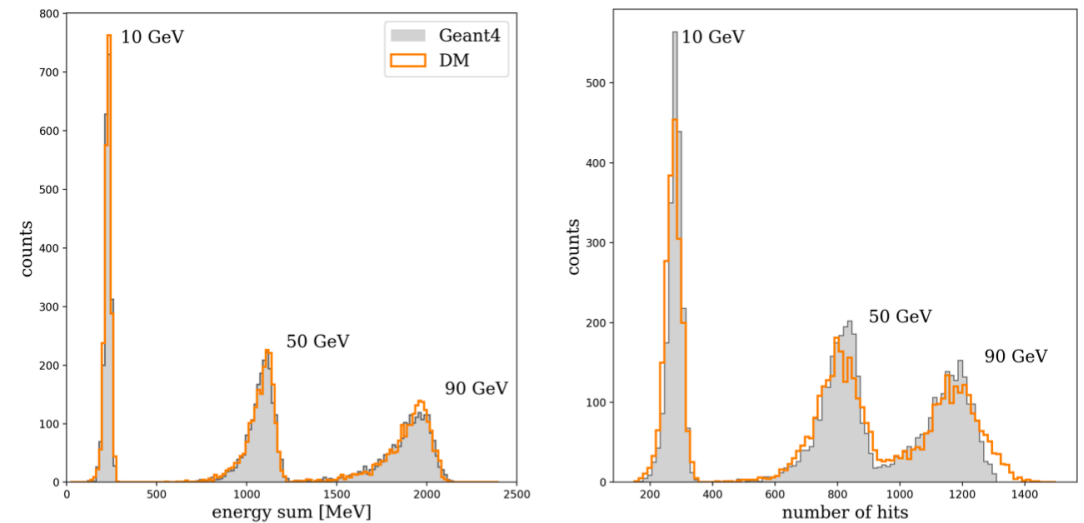
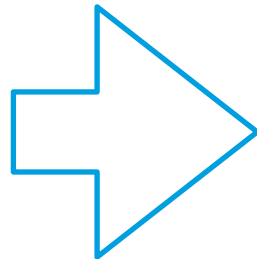
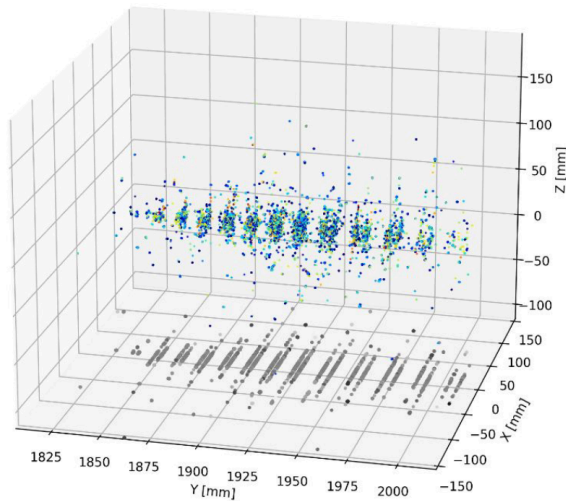
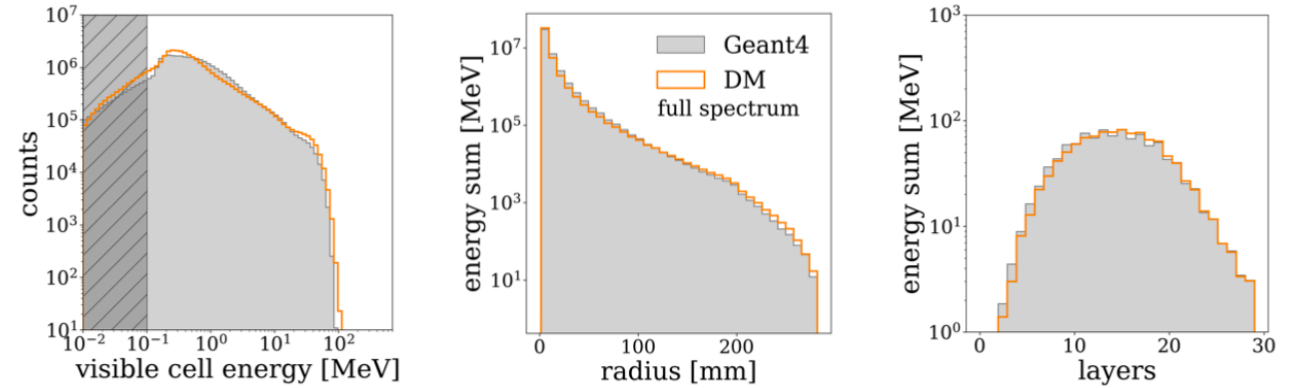
- recently good success with using **point clouds** and **diffusion models**
- using much higher granularity from Geant4 simulation to address issue with irregular cell geometries in 'real' calorimeters



Point Cloud Diffusion

A.Korol

- recently good success with using **point clouds** and **diffusion models**
 - using much higher granularity from Geant4 simulation to address issue with irregular cell geometries in ‘real’ calorimeters
- achieve promising first results
 - publication in preparation ...



ML inference in DDG4

running ML fast sim in 'real' simulations

- prototype library for running fast ML inference in DDG4/Geant4 based full simulation:
DDFastShowerML:
 - <https://gitlab.desy.de/ilcsoft/ddfastshowerml>
- goal to have an easy to use library that can easily be extended to all types of ML architectures for all detectors in DD4hep
- will use some time of the **hackathon discussion** to iterate this idea and see if this is the best way to integrate ML in DDG4/Geant4

MODEL

- detailed knowledge of ML Architecture

- prepare input
- convert output
- eg. GAN, BIB-AE, PointCloud

- concrete C++ inference, e.g.
- ONNX, libtorch, ...

INFERENCE

GEOMETRY

- concrete detector geometry, e.g.
- ILD Ecal barrel
- " " endcaps

ML inference in DDG4

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```
/// User callback to model the particle/energy shower - details defined in ML_MODEL
virtual void modelShower(const G4FastTrack& track, G4FastStep& step) override {

    // remove particle from further processing by G4
    step.KillPrimaryTrack();
    step.SetPrimaryTrackPathLength(0.0);
    G4double energy = track.GetPrimaryTrack()->GetKineticEnergy();
    step.SetTotalEnergyDeposited(energy);

    _input.clear() ;
    _output.clear() ;
    for( auto& layerSPs : _spacepoints )
        layerSPs.clear() ;

    fastsimML.model.prepareInput( track, _input , _output ) ;

    fastsimML.inference.runInference(_input, _output ) ;

    fastsimML.model.convertOutput( track, _output , _spacepoints ) ;

    fastsimML.geometry.localToGlobal( track, _spacepoints ) ;

    // now deposit energies in the detector using calculated global positions
    for( auto& layerSPs : _spacepoints )
        for( auto& sp : layerSPs ) {
            fastsimML.hitMaker->make( G4FastHit( G4ThreeVector(sp.X,sp.Y,sp.Z) , sp.E ), track);
        }
}
```

Summary and Outlook

- calo ML group in Hamburg (DESY/UHH) focuses on high fidelity generation of showers in highly granular calorimeters
- investigating different ML architectures and topics:
 - first **generation of pion showers** in high granular calorimeter - incl. **reconstruction**
 - **energy and angular conditioning** for EM showers with high fidelity
 - **point cloud diffusion** model for EM showers
 - started to develop *DDFastShowerML* library to run inference in DDG4
- continue work towards final goal: full calorimeter simulation w. ML (in Geant4) for ILD
 - test ML model with physics observables after reconstruction, e.g. di-boson mass or JER
 - continue work on combining ECal&HCal calorimeters for hadron simulation
 - ...