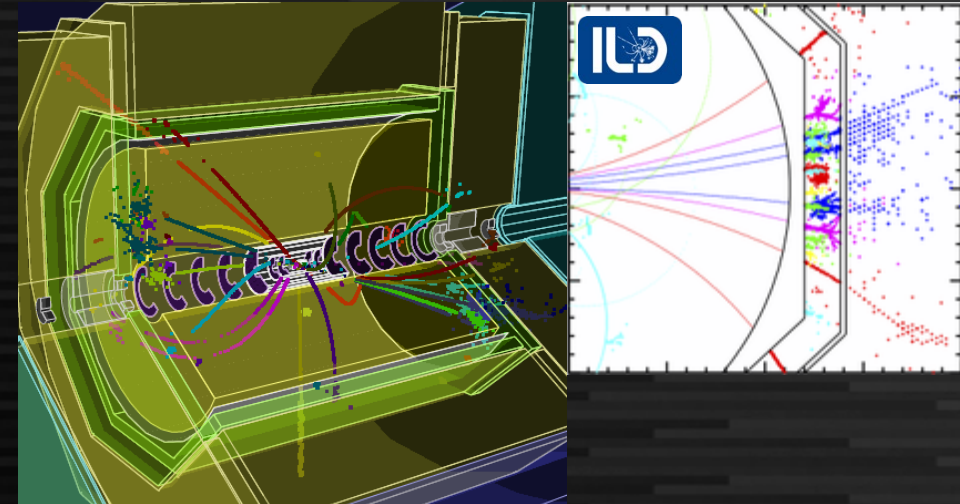


# Particle flow with GNN: progress report

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# Particle flow for Higgs factories

- High granular calorimetry
  - 3D pixels for imaging EM/hadron showers at calorimeters
    - eg.  $10^8$  channels for ILD ECAL
  - Separation of particles inside jets
    - $\sim 2x$  better energy resolution by separation of contribution from charged particles
      - Software algorithm essential (as well as hardware design)
- Particle Flow algorithm
  - Essential algorithm for high granular calorimetry
  - Complicated pattern recognition



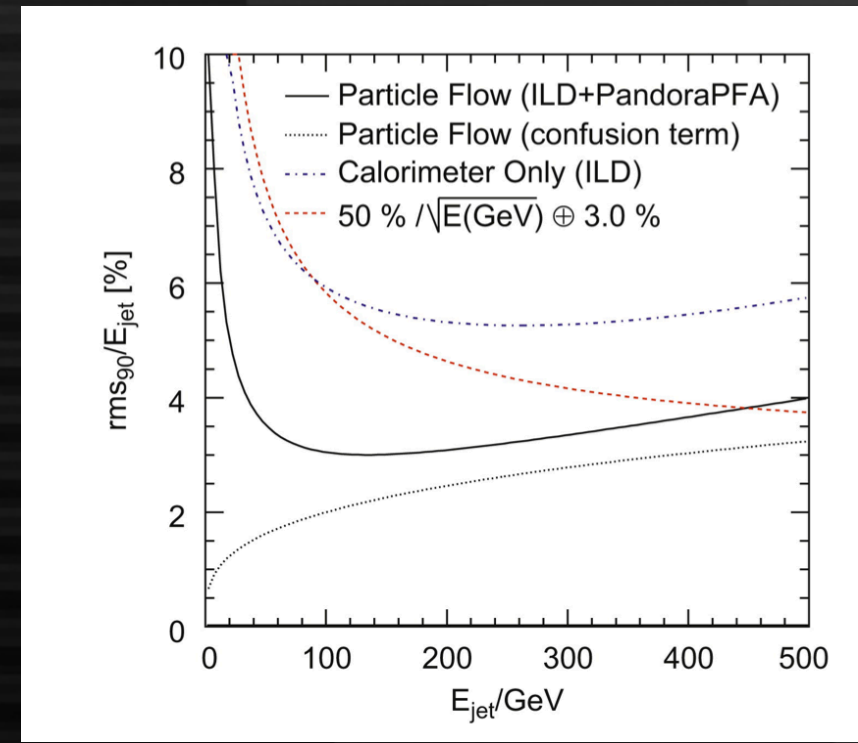
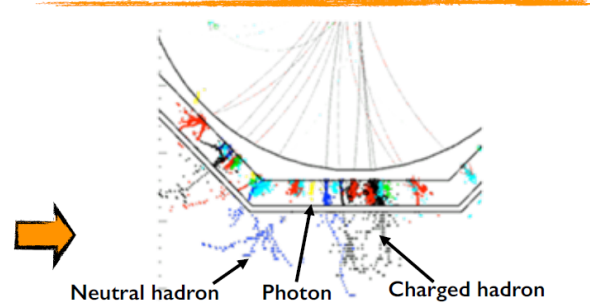
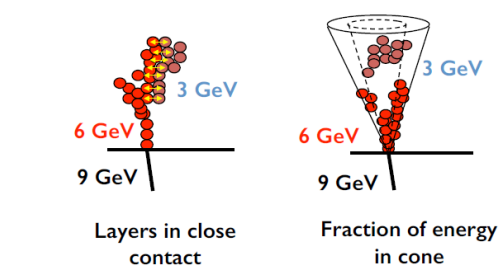
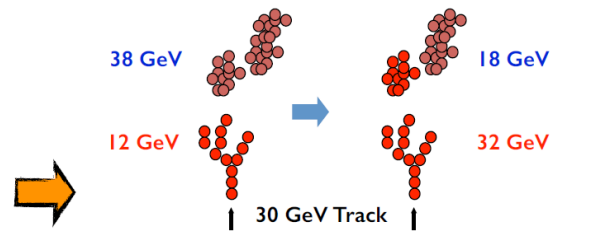
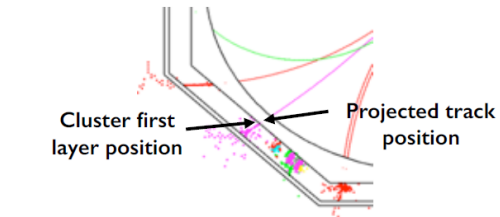
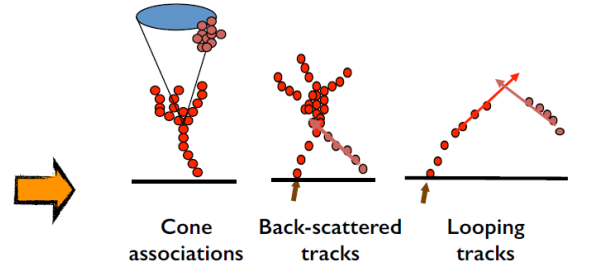
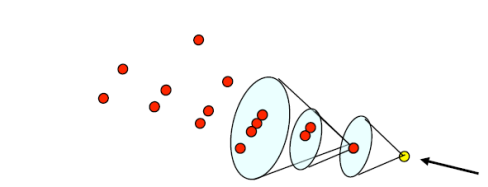
# Pandora ParticleFlow algorithm

## Pandora LC Algorithms



60+ algorithms for fine-granularity detectors

- ConeClustering Algorithm
- Topological Association Algorithms
- Track-Cluster Association Algorithms
- Reclustering Algorithms
- Fragment Removal Algorithms
- PFO Construction Algorithms



Widely used since 2008  
Reasonably good performance up to ~50 GeV jets  
Confusion dominates at higher energies

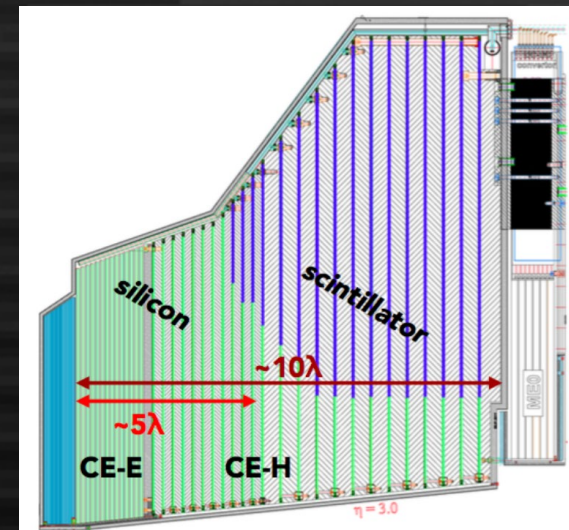
# Motivations for DNN particle flow

- Performance improvement
  - Confusion dominant at jet energy  $> 100$  GeV
  - More efficient way to separate cluster from charged particles should be investigated
- Integrate other functions
  - Software compensation, particle ID etc. closely related to PFA
- Detector optimization
  - Comparison with different detector settings
    - PandoraPFA too much depends on internal parameters
  - Effect of timing information to be investigated
    - With different timing resolution (1 ns, 100 ps, 10 ps, ...)

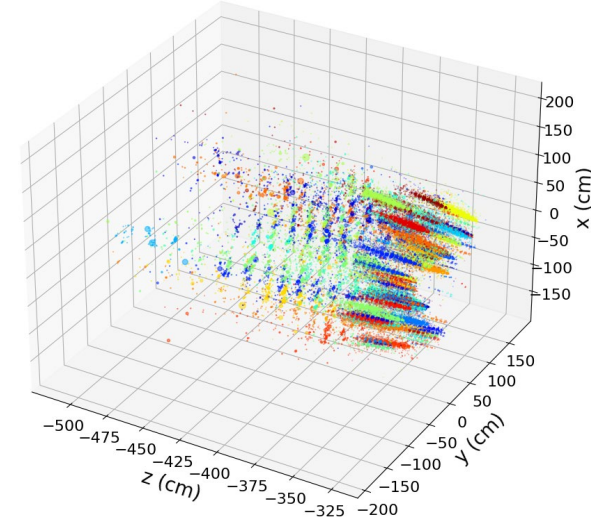


# GravNet for CMS HGCAL

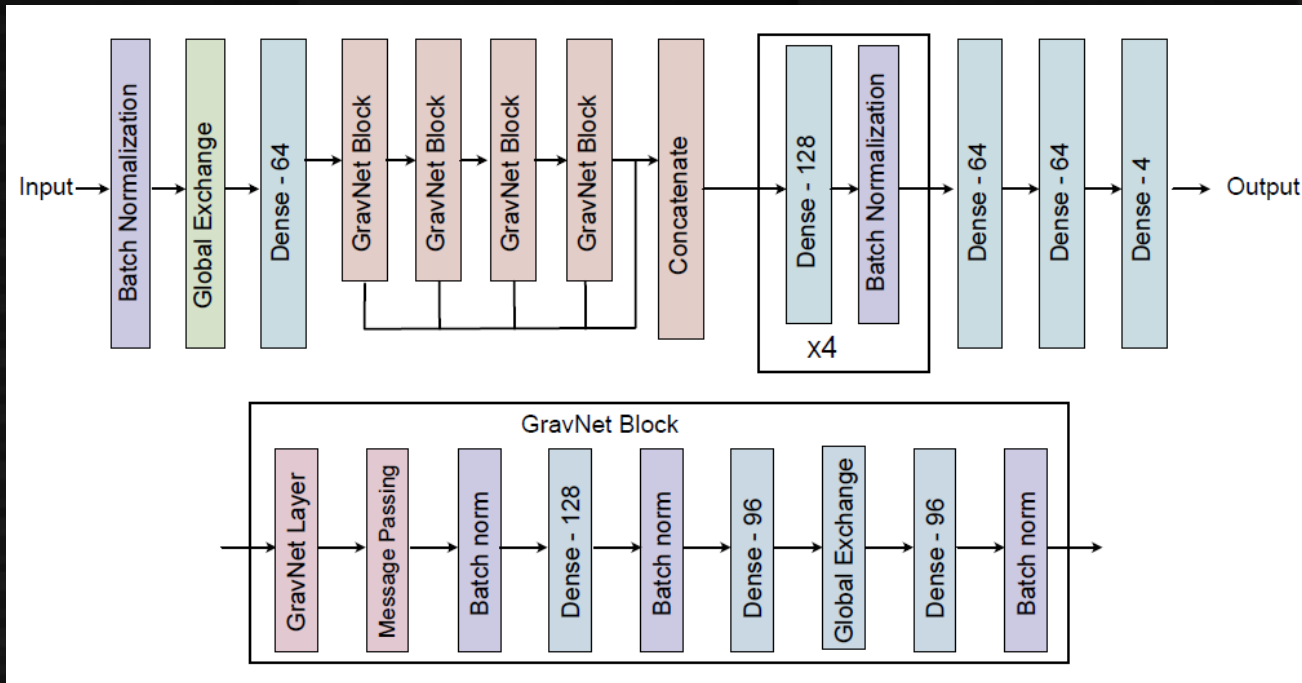
- CMS HGCAL
  - High granular forward calorimeter for HL-LHC upgrade at CMS
  - Similar to ILD calorimeter (silicon pixel + scintillator)
    - Inspired by CALICE development
- Reconstruction at HGCAL
  - Big noise to be separated by software
  - Numerous particles from  $\sim 200$  pileups
    - Difficult to handle by current software
    - DNN reconstruction being investigated
      - Reasonable performance obtained up to  $\sim 50$  pileups?



CMS Phase-2 Simulation Preliminary



# The network



Rather complicated network with ~30 hidden layers

“Object condensation” loss function is applied (shown in next page)

**Input/output obtained for each hit at calorimeter**

Input: Features at each hit (position, energy deposit, timing)

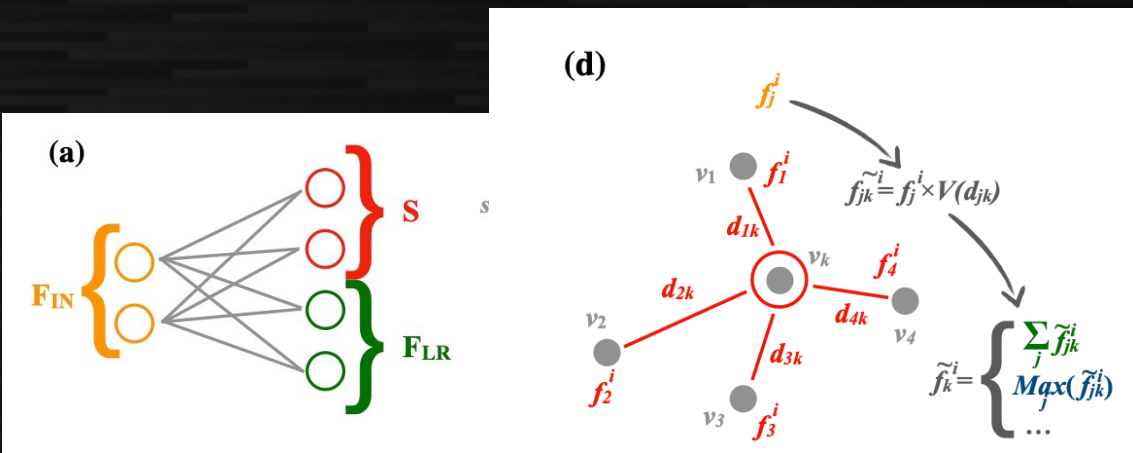
Output: “condensation coefficient”  $\beta$ , position at virtual coordinate (2-dim)  
optional output of features such as energy, PID (not used now)

Dense (fully-connected layer) inside each hit, GravNet connects hits

# GravNet and Object Condensation

GravNet arXiv:1902.07987

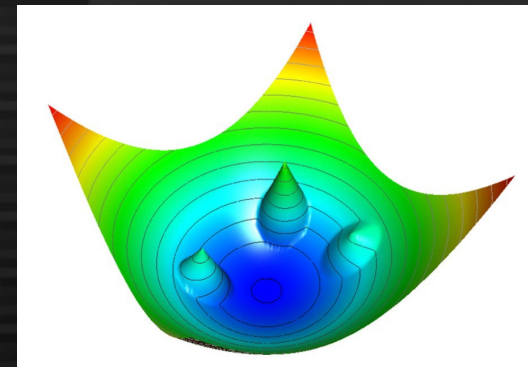
- The virtual coordinate (S) is derived from input variables with simple MLP
- Convolution using “distance” at S (bigger convolution with nearer hits)
- Repeat 2 times and concatenate the output with simple MLP



Object Condensation (loss function)

$$L = L_p + s_C(L_\beta + L_V)$$

arXiv:2002.03605



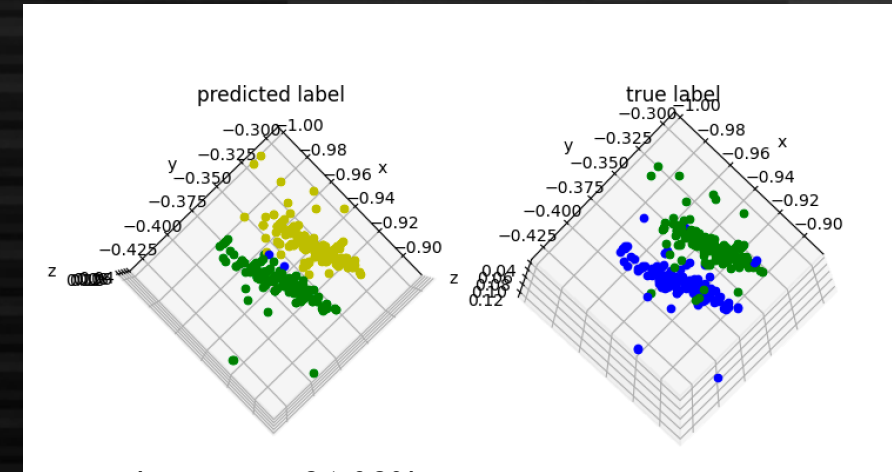
- **Condensation point:** The hit with largest  $\beta$  at each (MC) cluster
- $L_V$ : **Attractive potential** to the condensation point of the **same cluster** and **repulsive potential** to the condensation point of **different clusters**
- $L_\beta$ : Pulling up  $\beta$  of the condensation point
- $L_p$ : Regression to output features (energy etc.)  $\rightarrow$  currently not used



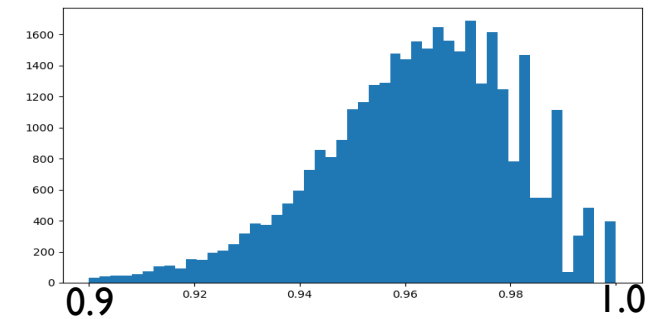
# Importing to ILD full simulation

- Prepare features from ILD full simulation
  - With recent versions (> v02-02)
- Input features: (x, y, z, edep)
- True cluster info from MCParticle and LCRelation
- Produced events
  - Two photons (5/10 GeV, fixed opening angles)
  - (n x ) taus (5/10 GeV)
- Evaluation
  - Fraction of hits associated to the correct cluster (accuracy)

Example of a two-photon event (5 GeV, 30 mrad)



Average = 96.08%



Reasonable performance seen

accuracy

Angle[mrad]	30	60	90	120	150
Accuracy[%]	96.08	98.64	99.30	99.68	99.56

For details, refer eg. <https://indico.slac.stanford.edu/event/7467/contributions/5948/attachments/2887/8032/230517-lcws2023-hlreco-suehara.pdf>

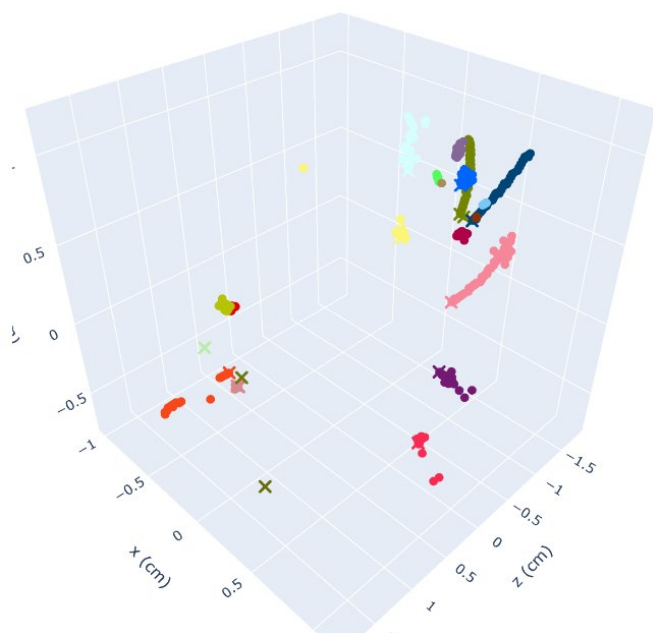


# Work in Progress: track-cluster matching

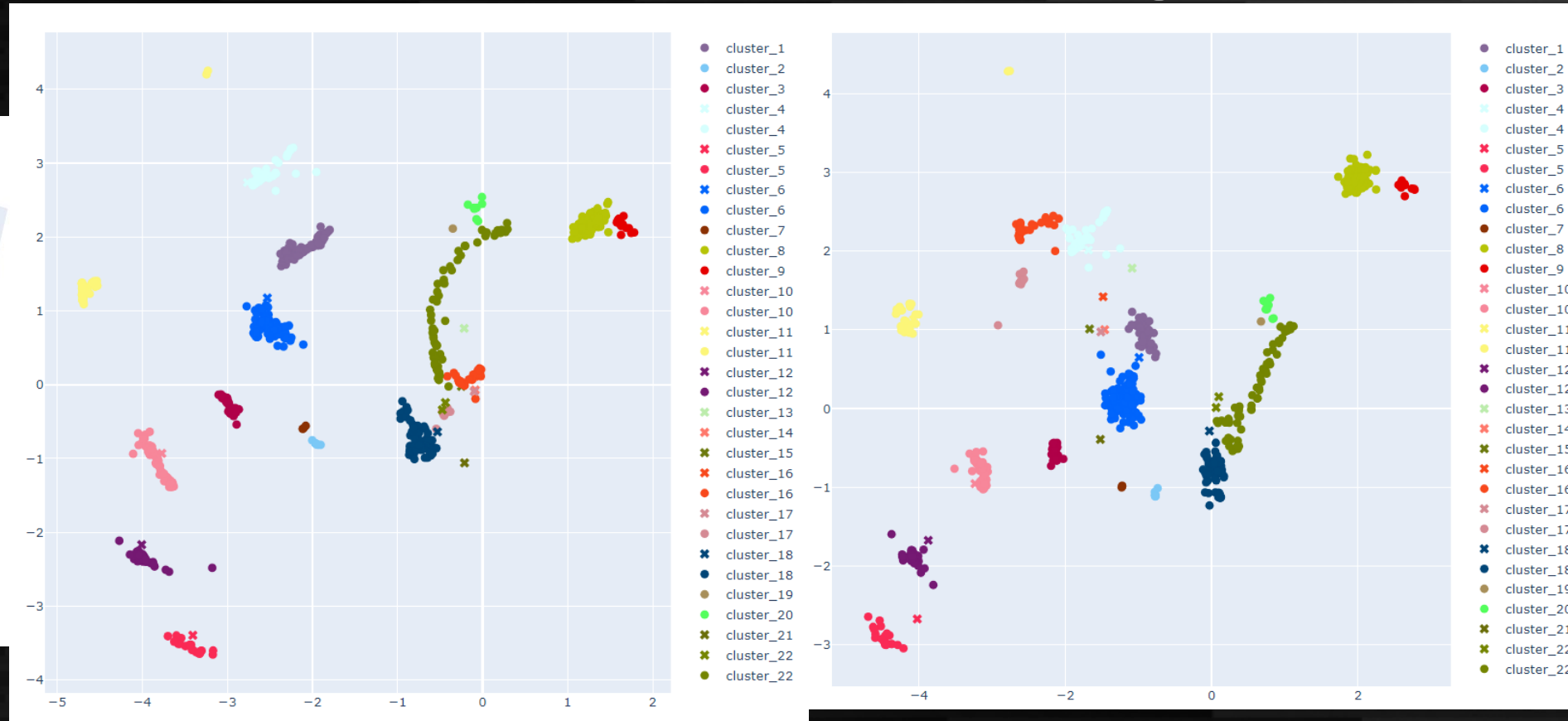
- PFA is essentially a problem “to subtract hits from tracks”
- HGICAL algorithm does not utilize track information
  - Only calorimeter clustering exists
- Simple extension to include track information
  - Adding “virtual hits” derived from track information
    - Hits at position where the track enters the calorimeter (from LCIO StackState)
  - Add a term to the object condensation loss function
    - Pulling up  $\beta$  of tracks (virtual hits) to promote them to condensation points (in addition to the usual beta-term, called **beta-track term**)
  - Evaluate fraction of (MC) charged clusters to be correctly assigned to clusters with tracks (virtual hits)

# Preliminary results – event sample

10 Taus @ 10 GeV each



Real 3D coordinate



Hits on the virtual coordinate – colored by MC truth clusters  
x refers virtual hits from tracks  
left with beta-track term, right without beta-track term

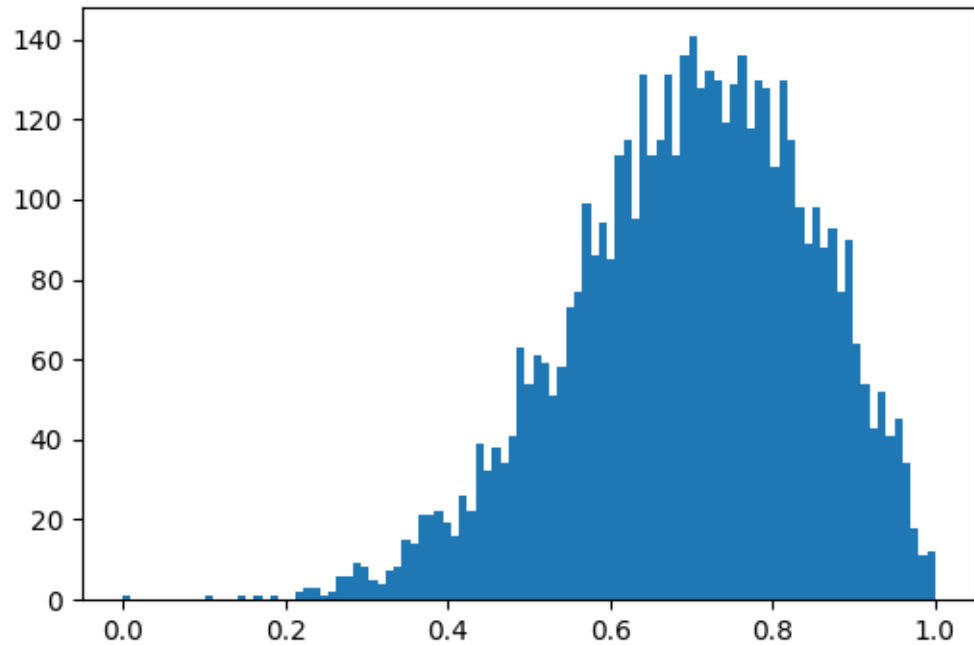
# Situations

- Position of virtual hit (from track) is distant from other hits from the same cluster even if we put beta\_track term
  - Better in initial 3D coordinate
  - Maybe mapping 0/1 condition (track=1, non-track=0) to the output coordinates then getting different position
  - Beta\_term only from epoch 8: is this the problem?
  - Better representation of beta\_track term
    - Being tried

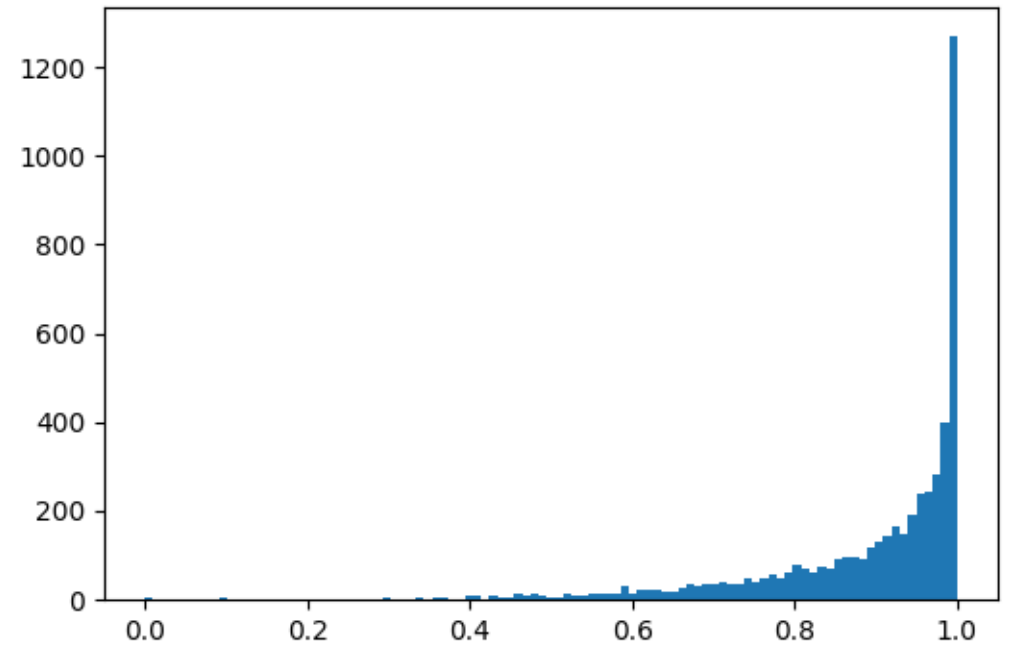
# Evaluation of performance

Fraction of charged hit predicted correctly

Fraction of neutral hit predicted correctly



Average: 0.698 (somehow low)



Average: 0.890

Pred charged hit: associated cluster having at least one track



# Ideas of improvement(short term / ongoing)

- Adding theta/phi → not significant
- Utilizing track momentum
  - At clustering (additional conditional association, being tried)
  - Regression of momentum at the network (not done yet)
- Adding classification of track hit or not (being done)
  - Interaction/tuning with other terms
- Tuning of learning (partially done, not significant so far)
- Fully-DNN clustering (currently clustering is done with traditional way)

# Summary / long-term plans

- New DNN-based particle flow algorithm is under development based on clustering at CMS HGCAL study
- Track-cluster matching is being implemented, statistical results will come soon
  - Energy regression with track momentum information will be the next step of implementation
- Medium/long term plans (or just hopes)
  - Can be extended to any analyses using cluster/jet information using the PFA as “a foundation model”
    - Such as Particle ID, Jet clustering, even physics analyses directly
  - “Differentiate” detector parameters/designs for optimization

