

A fast convolutional neural network for identifying LLP decays in a high-granularity calorimeter

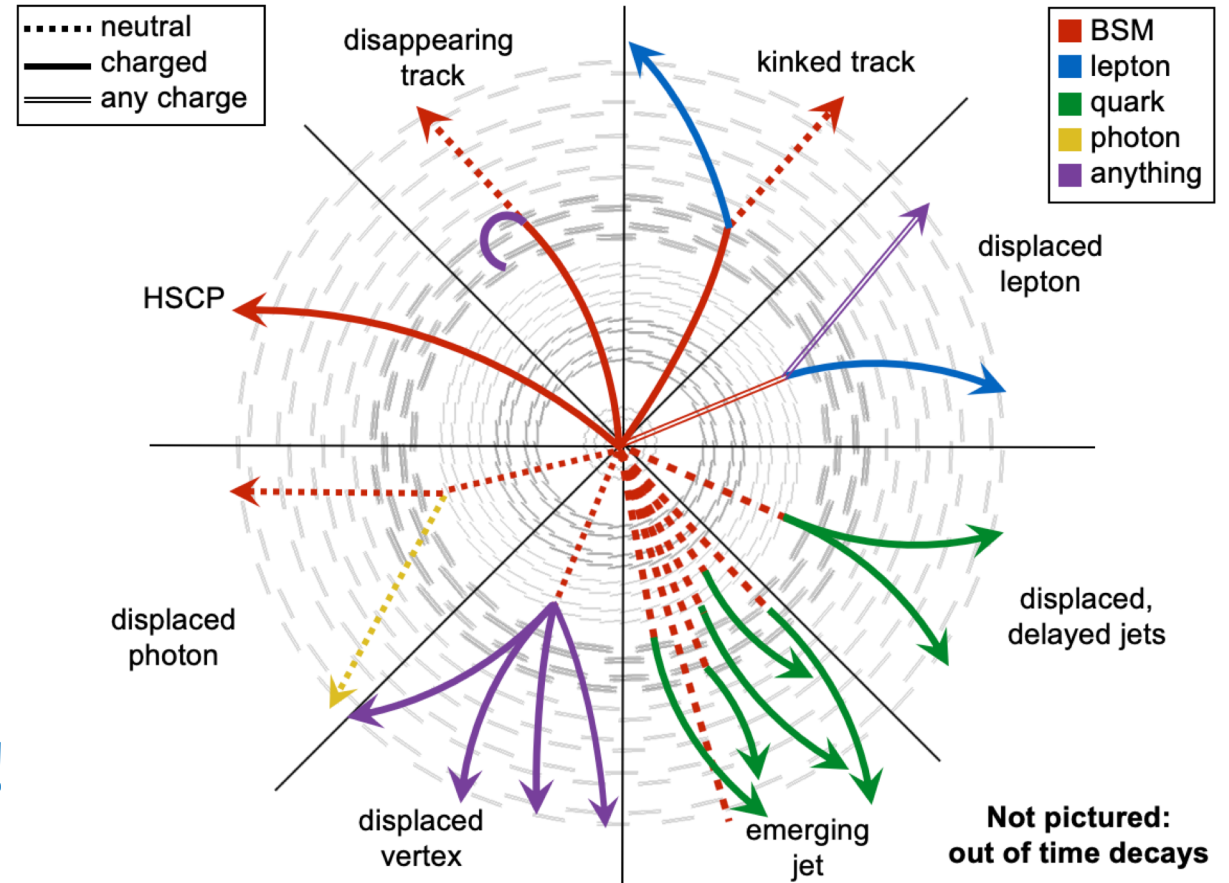
Juliette Alimena, Yutaro Iiyama, Jan Kieseler

LLP8

November 19, 2020

Triggers for Long-Lived Particles

- LLP searches span a wide variety of signatures, models, lifetimes, masses, decay locations, etc.
- The signatures are often **unusual** and **not covered** by “standard” reconstruction or triggers
- If your data is not triggered, it’s lost!
- **Dedicated triggers for LLPs are crucial!**



Displaced Particles in CMS Phase 2 at Level 1

- **Can we trigger at level 1 (L1) on displaced/ delayed particles in the CMS at the HL-LHC?**

- Track trigger:

- Only for charged particles
- Only for $|\eta| < 2.1$
- Only for $|d_0| < 10$ cm, with the track trigger extension

- MIP Timing Detector (MTD):

- Will not be used at L1

- ECal and HCal barrel:

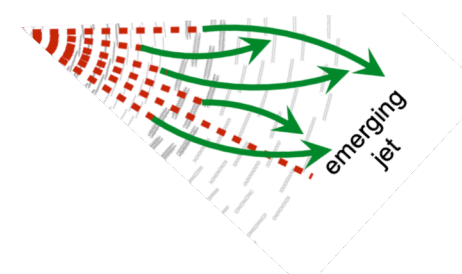
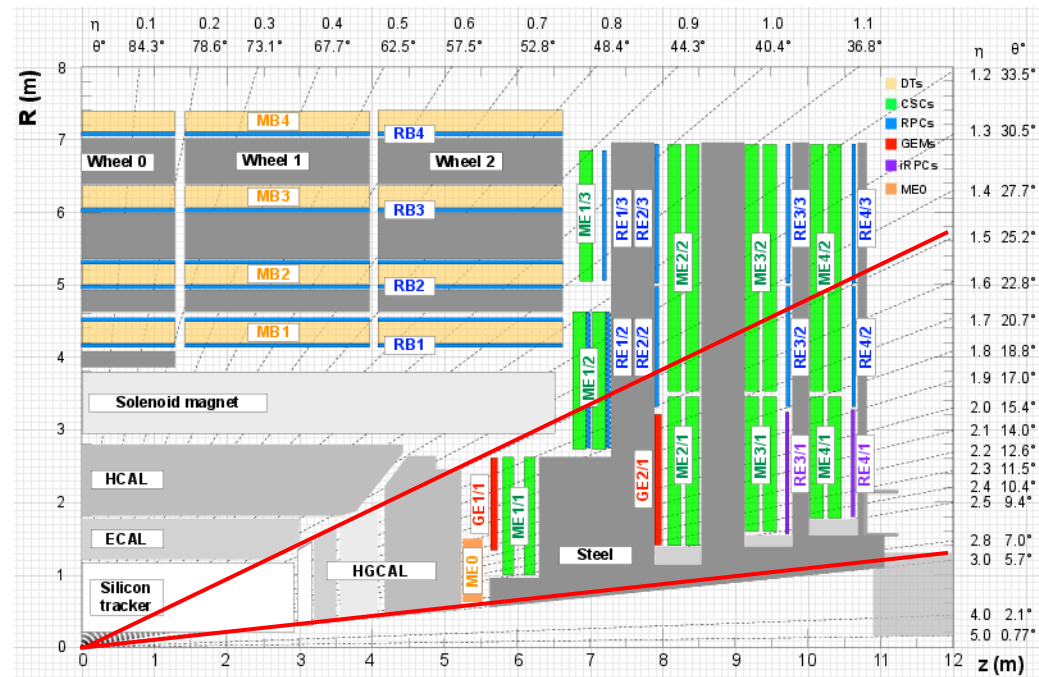
- Timing available at L1

- High-Granularity Calorimeter (HGCAL):

- No timing at L1
- Current HGCAL L1 reconstruction assumes pointing showers

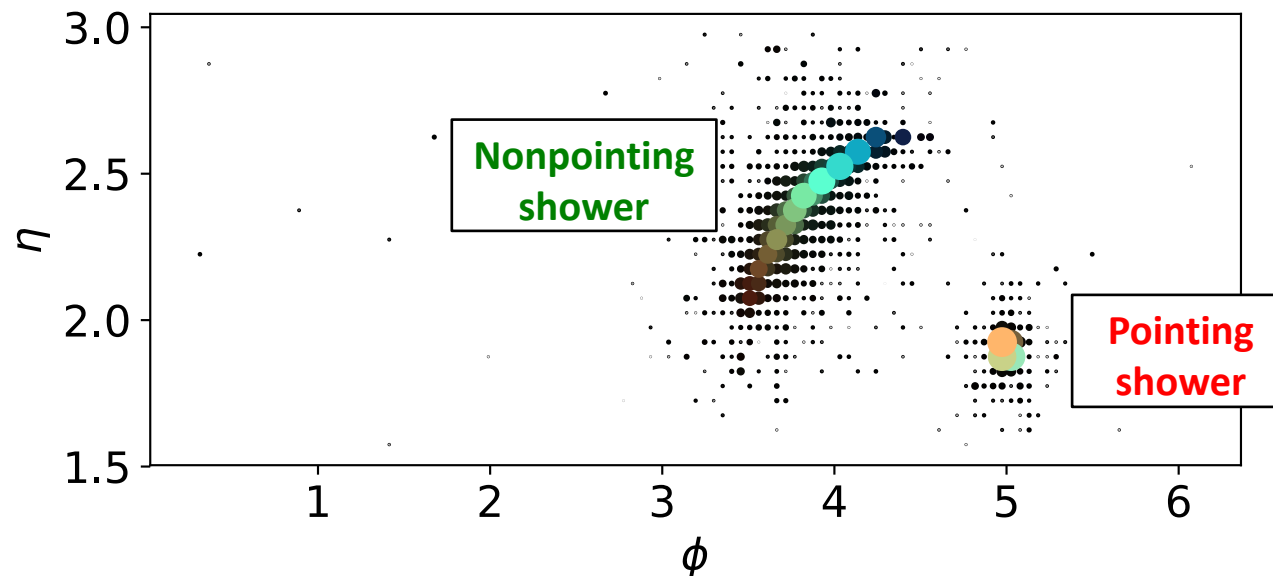
- **Displaced/delayed signatures from LLPs in the forward region could be completely missed with CMS in Phase 2, without a dedicated trigger**

- Example LLP signature: Emerging jets (t-channel production)



CNN Trigger for LLP Decays in HGCal

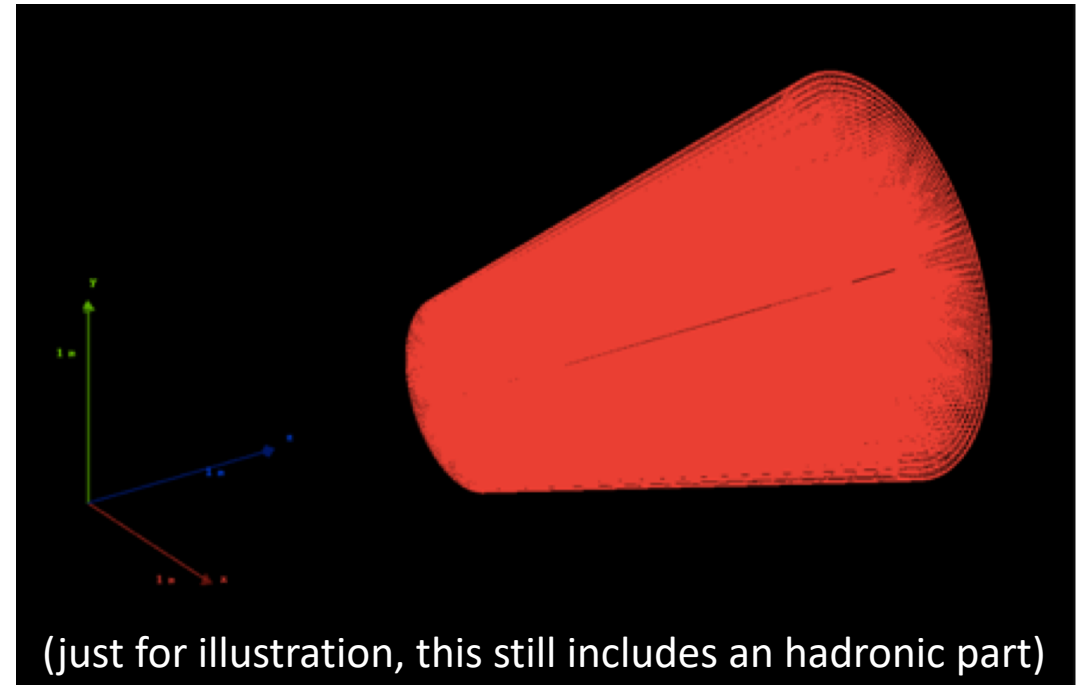
- Developed a **fast convolutional neural network** (CNN) to find **nonpointing showers** in a **high-granularity calorimeter** (HGCal)
- Computer vision image recognition can easily differentiate between nonpointing and pointing showers
- Proof of concept paper (<https://arxiv.org/abs/2004.10744>) recently accepted by JINST



Colors indicate calorimeter layer number
Marker size indicates deposited energy

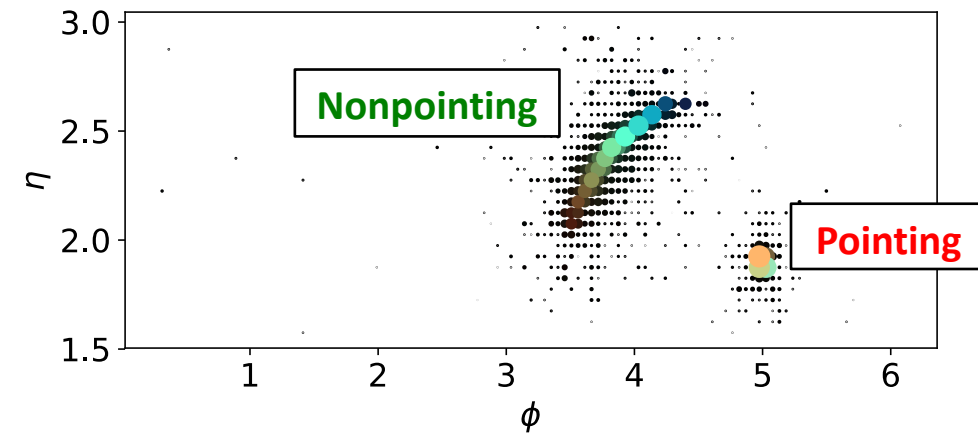
Detector

- Toy calorimeter in Geant4
- Granularity, geometry, and material like the EM part of CMS HGCal at L1 trigger
- Use projective geometry, regular η - ϕ grid
 - 30 x 120 x 14 pixels in η/ϕ /layer
 - NB: CMS HGCal will have sets of hexagonal sensors of different sizes: will need to make this structure fit in a grid, in order to use it with CNNs
- 1T magnetic field



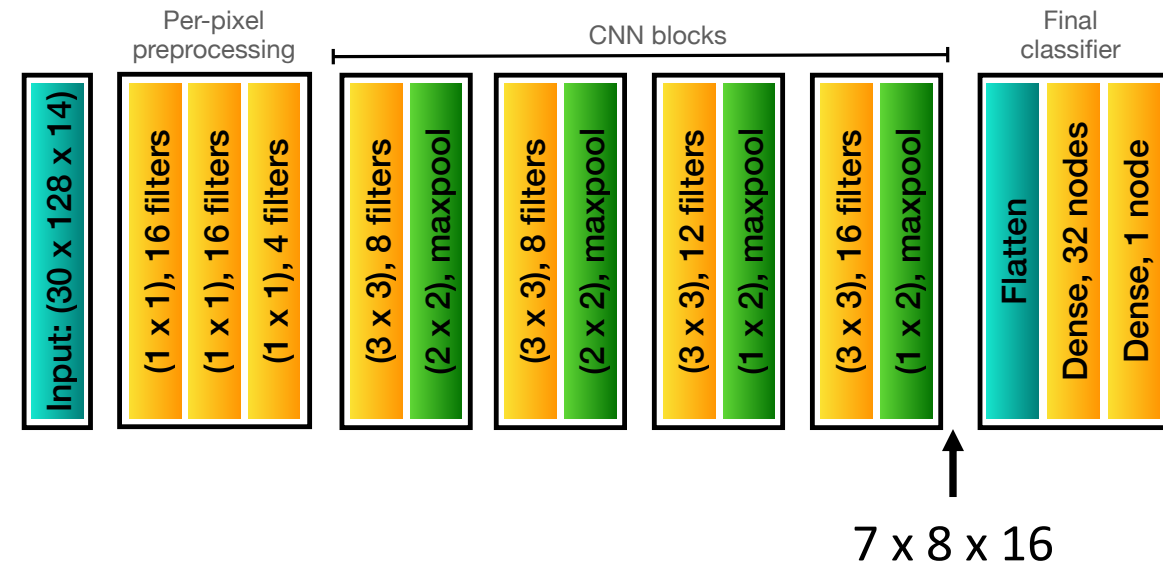
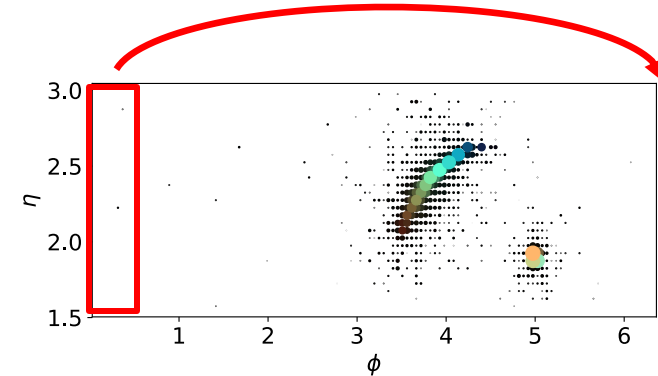
Data Sample

- Signal:
 - Focus on EM showers
 - Generate photons in Geant directly in front of calorimeter, with displacements between 20 and 60 cm
 - Choose angle wrt projection axis (α) uniformly between 0 and $\pi/3$
 - Adjust rotation such that particles traverse the calorimeter
 - Select energy between 10 and 200 GeV
 - 780k events for training, 8.8k for validation, 14.4k for testing
 - **Model-independent**
- Minimum-bias background and pileup:
 - CMS minbias configuration
 - Generate individual events (15.3M for training, 4M for testing and validation) with Pythia8
 - Mix 200 randomly selected events to form one minbias event
 - Overlay with signal

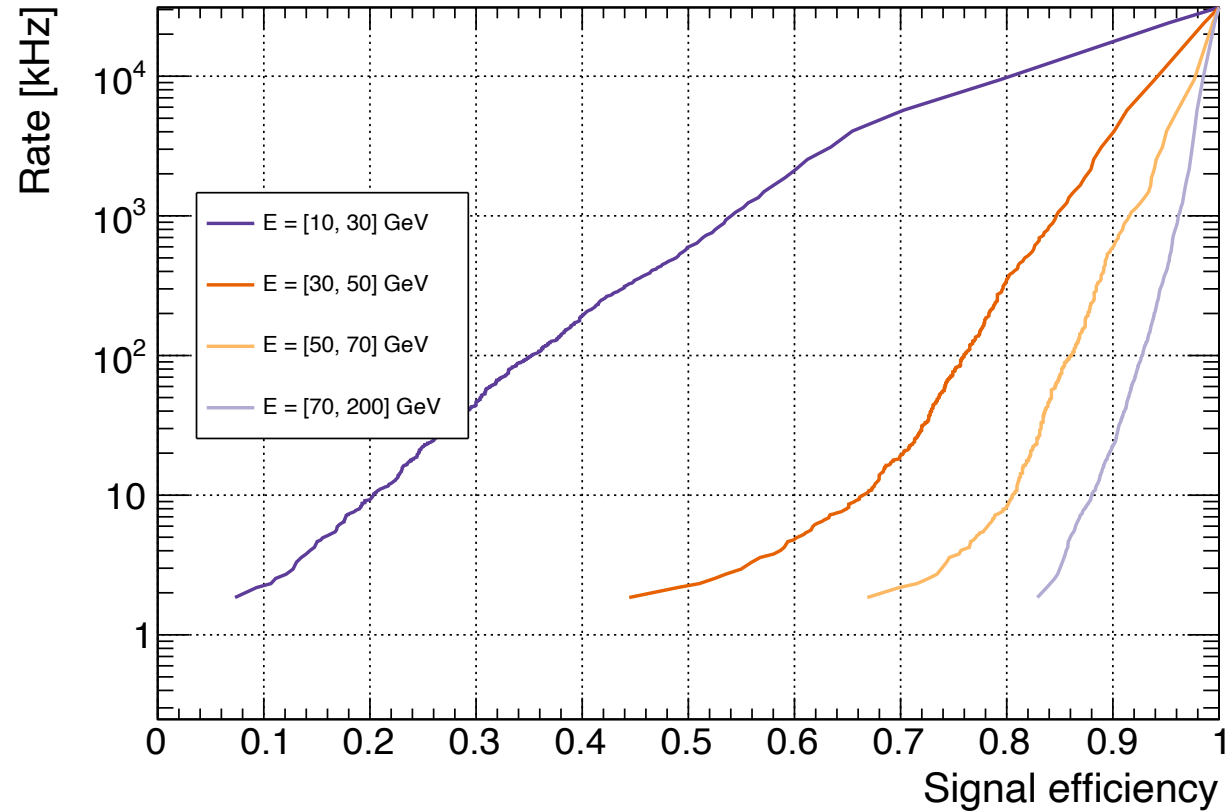


Neural Network and Training

- **Simple CNN designed to provide compromise between performance and resource requirements**
- Pixels from $\phi=0$ to $\phi=0.4$ are repeated at $\phi=2\pi$ to account for particles that enter the calorimeter at $\phi \sim 0$
 - Therefore 120 pixels in ϕ become 128
- Per-pixel preprocessing: 14 layers \rightarrow 4 “colors”, to make input to CNN blocks small
- Simple CNN + max pool blocks
- Final classifier with low parameter count
- Trained with Adam optimizer
 - 1 epoch with a batch size of 50, learning rate of $1e-4$
 - 30 epochs with a batch size of 500, learning rate of $3e-4$
 - 1 signal event : 1 minbias event for training
- 1 signal event : 70 minbias events for testing



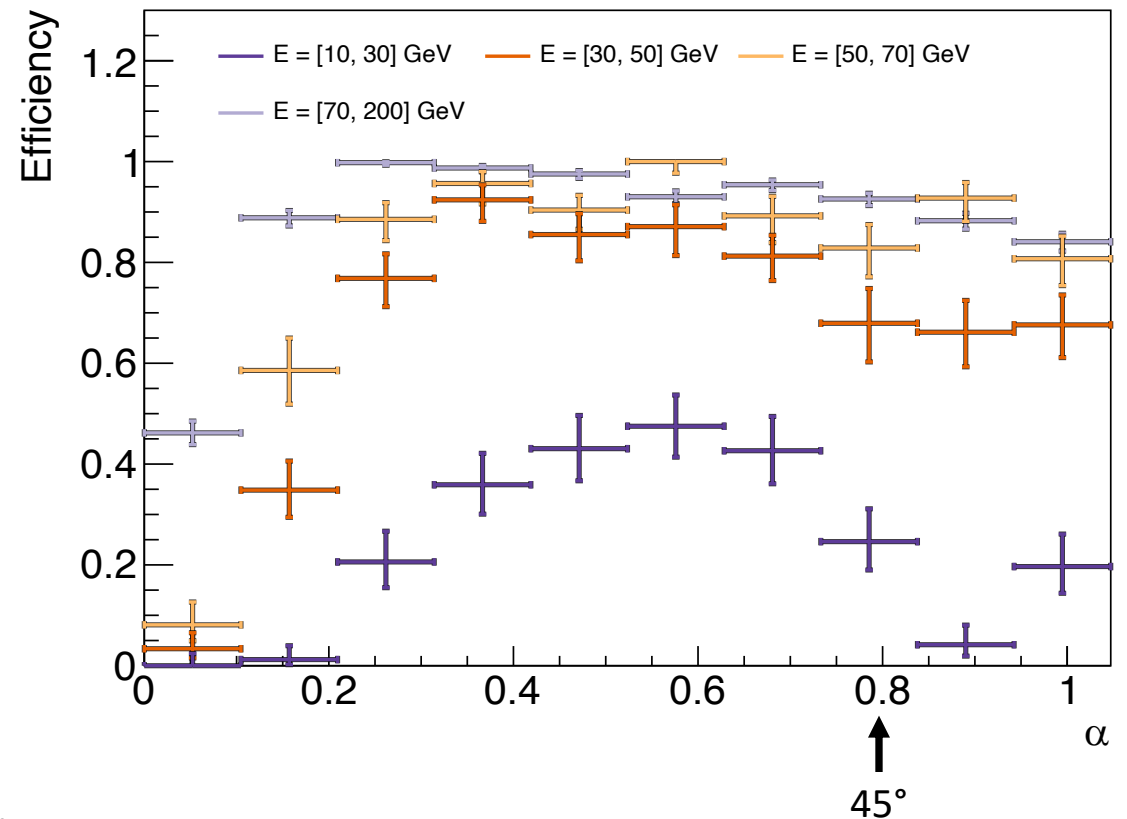
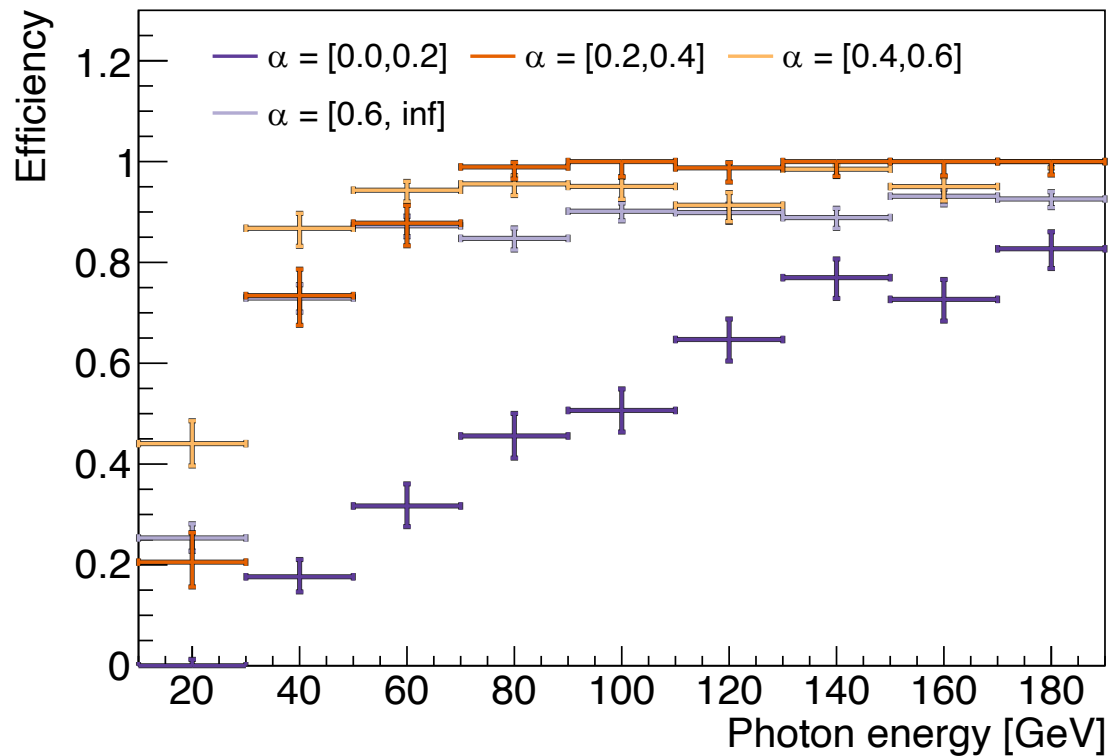
Results



Good performance already at low energy (this is energy, not p_T)

Efficiency vs Energy and Angle

- Select a working point of 15 kHz based on previous slide
- Promising, model-independent results



Summary and Outlook

- Showed that a fast CNN-based pattern recognition trigger is able to identify nonpointing showers in a forward, high-granularity calorimeter with reasonable efficiency and rate
- Dependence on energy and shower angle behaves as expected
- Now working with CERN HLS4ML group to see how feasible this would be for FPGAs in the real Phase 2 CMS HGCal at L1
- Ultimate goal is to have this as a trigger for the CMS HGCal