

Machine Learning Approaches to Particle Identification in the DUNE Far Detector

Isobel Mawby (they/them) for the DUNE collaboration



email: i.mawby1@lancaster.ac.uk

1 Overview

- 1) Developments of the **Pandora reconstruction software** **3** should be motivated by **analyses**
- 2) We target the search for **CP-violation with the DUNE far detectors** **2**
- 3) Here one identifies **charged current (CC) ν_e/ν_μ interactions**
- 4) In the **Pandora-based analysis**, **4** the success of this relies on:
 - a) a performant reconstruction software
 - b) excellent particle identification (PID)**
- 5) Here we focus on a machine learning approach to particle identification **5**

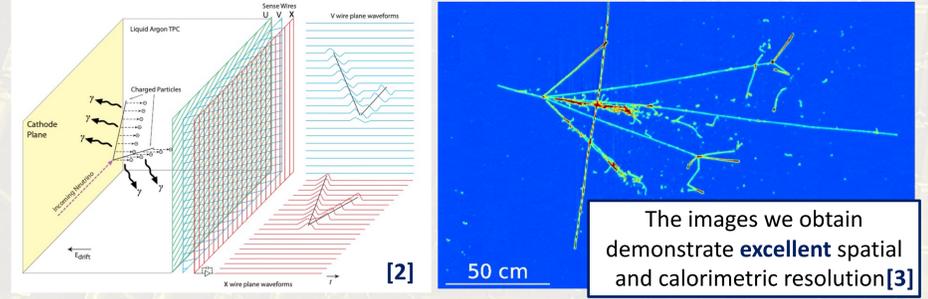
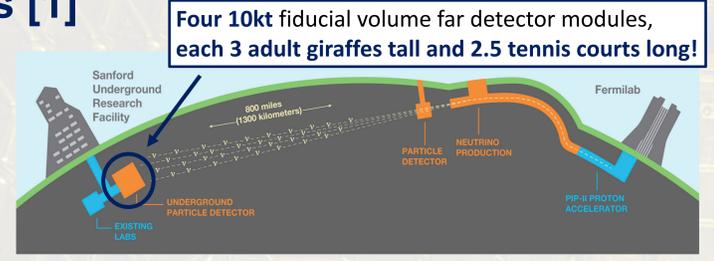
2 DUNE and its Far Detectors [1]

DUNE primarily aims to:

- Make **precise neutrino oscillation measurements**
- Search for **BSM physics** (e.g. proton decay)
- Study **low energy neutrinos** (e.g. supernova neutrinos)

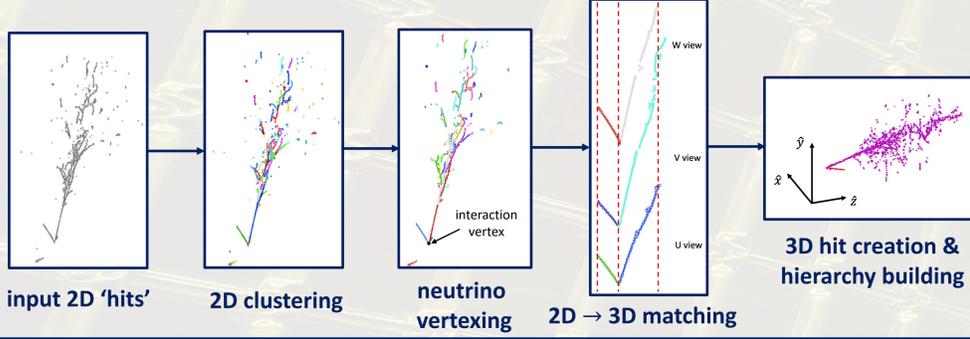
Three modules will be **Liquid-Argon Time-Projection Chambers**:

- Neutrinos interact with **LAr nuclei**
- Light collected by **photon detection system**, providing timing information
- Outgoing charged particles **ionise the LAr nuclei**
- Ionisation electrons drifted to **readout planes by an applied electric field**



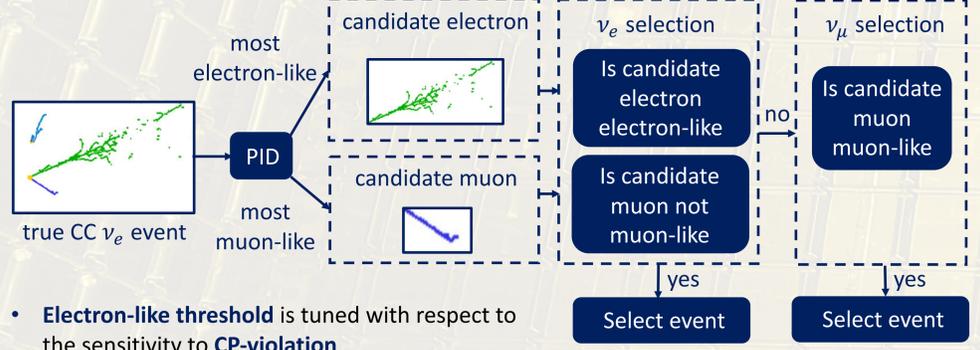
3 Pandora – the hope left in the jar [4,5]

- Pandora is a **pattern recognition software**, used to reconstruct neutrino interactions
- The detail of the fine-grain images (see **2**) presents a huge reconstruction challenge
- Pandora overcomes this with a **'multi-algorithm approach'**, where the reconstruction is split into stages composed of many 'hand engineered' and machine learning algorithms



4 The Pandora-based CC ν_e/ν_μ selection [6]

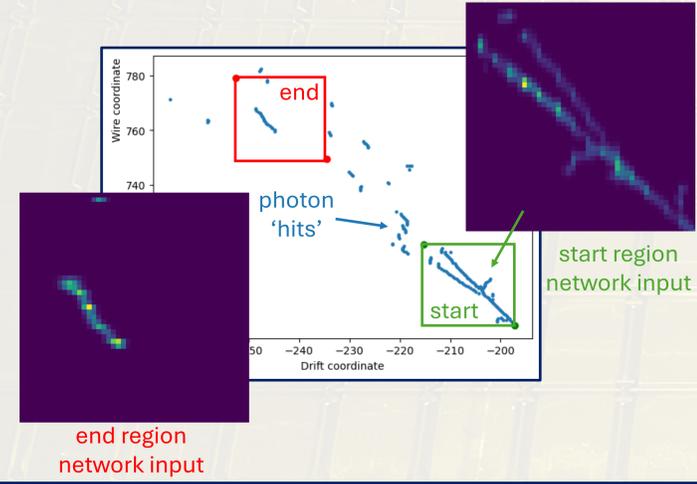
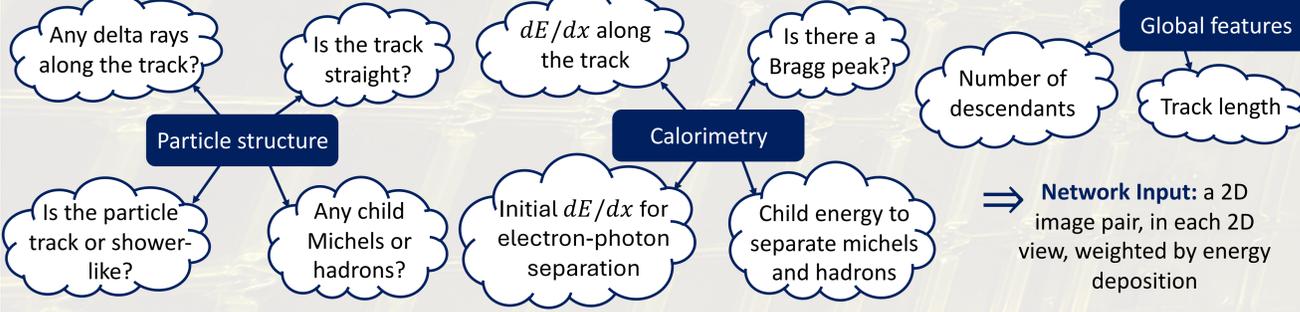
- Events are selected based on the assumed identity of the leading lepton (if it exists)



- **Electron-like threshold** is tuned with respect to the sensitivity to **CP-violation**
- **Muon-like threshold** is tuned with respect to the product of the **CC ν_μ selection efficiency and purity**

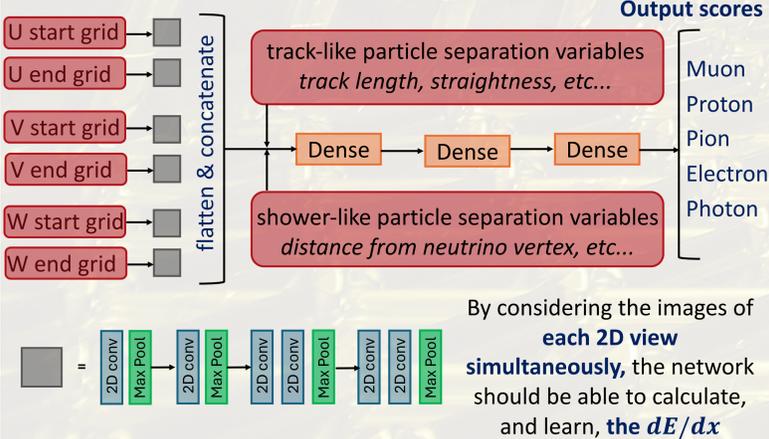
5 Convolutional Neural Network for PID

- One of the most important choices when designing any network is the **form of the input**
- So, what do we investigate when trying to identify a given particle?



6 Network Architecture

- Architecture is a composition of **2D convolution** and **max pooling** layers
- Output particle type probability scores sum to **unity**

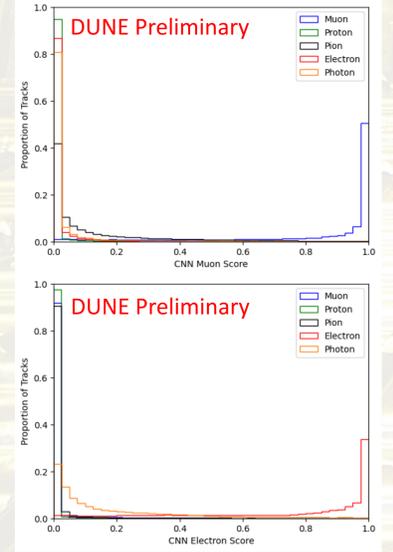
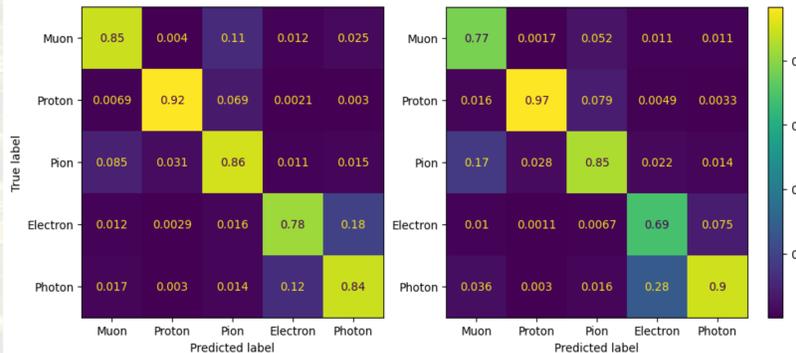


By considering the images of **each 2D view** **simultaneously**, the network should be able to calculate, and learn, the **dE/dx**

7 Results

- The convolution neural network approach achieves **good particle separation**
- As expected, main confusion between **π/μ and e/γ**

Efficiency normalised (read horizontally) Purity normalised (read vertically)



DUNE Preliminary	Selection Efficiency	Selection Purity
ν_e Selection	67.7%	72.7%
ν_μ Selection	92.8%	93.2%

8 What's next?

- The presented analysis has **directly motivated** several areas of ongoing developments to the Pandora reconstruction, most of which focus on the **reconstruction of electrons and photons**
- The use of **innovative machine learning** in the multi-algorithm approach is being investigated and is found to be performant

References

- [1] DUNE Collaboration. (2020) JINST 15(08):08008
- [2] DUNE Collaboration. (2020) JINST 15(08):08010
- [3] DUNE Collaboration. (2020) JINST 15(12):12004
- [4] MicroBooNE Collaboration. (2018) Eur Phys J C 78(1):82
- [5] DUNE Collaboration. (2023) Eur Phys J C 83(7):618
- [6] Isobel Mawby. (2023) FERMILAB-THESIS-2023-11

- The **ν_e selection** is believed to be limited by the **reconstruction of electron showers**, particularly their **contamination and incomplete growth**
- The **ν_μ selection** is believed to be limited by **π/μ confusion**