High Energy **P**article **ID**entification **(PID)** —— An update on what I've been working on so far ———————

Quach Christine

Outline.

1. Setting the Context: SK/HK experiments and Existing PID Algorithms 2. My Work: Adapting Algorithms for High Energy Neutrinos

3. What's next?

A. Super-Kamiokande: Operational Principles and Observed Neutrinos

B. Current State of PID Algorithms

Setting the Context

- 41.4 m in height and 39.3 m in diameter, which holds approximately 50 ktons of ultrapure water.
- 11,146 PMTs

A. Super-Kamiokande: B. Current State of PID Algorithms

Setting the Experiments

Context

Super-Kamiokande

Hyper-Kamiokande

- An order of magnitude bigger than SK,
- 71 m in height and a diameter of 68 m
- 20 000 ultra-high sensitivity PMTs

Setting the **Observed neutrinos**

Relic Supernova neutrinos

- Neutrinos were produced in a supernova that occurred in the distant past and is still traveling through the universe today.
- # of events in SK: 10/year.
- Energy: 2 MeV (LOW)
- Provide a unique opportunity to study the properties of supernovae and the physics of the early universe

Neutrinos produced in the Sun by nuclear fusion . reactions. Vast majority of neutrinos passing through the

Solar neutrinos

- Neutrinos coming from a supernova explosion
- Energy ranging from 10 to 30 MeV. (HIGH)
- Interesting particles to observe to understand the p h y s i c s b e h i n d t h e explosion as neutrinos hold 99 % of the information about the explosion.
- Earth
- Most of solar neutrinos have energy below 10 MeV
	- Representation of the flux as a function of the energy of solar neutrinos according to the Standard Model of the Sun

A. Super-Kamiokande: B. Current State of PID Algorithms

Transient Supernova neutrinos

Setting the **Neutrino detection method**

Context

A. Super-Kamiokande: B. Current State of PID Algorithms

Cherenkov light

- The neutrino that interacts with electrons or nuclei in water produces a charged particle moving faster than the speed of light in water, which is slower than the speed of light in a vacuum.
- A cone of light is formed as a result, which is known as Cherenkov radiation.
- **.** The equivalent in the optical field is the sonic boom. PMTs record the Cherenkov light projected as a ring on the wall of the detector.

Neutrino detection method

1 A. Super-Kamiokande: B. Current State of PID Algorithms

Muon (Sharp) Electron (Blurry)

Merci Antoiline

Setting the Low energy

Why a Graph Neural Network (GNN)?

- - **Use of Deep Learning** ۳.
- \triangleright Why not a Convolutional <u>N</u>eural <u>N</u>etwork (used for image analysis)?
	- ۳. (instead of one small graph for a GNN)
	- ÷,

 \Rightarrow Smaller dataset and faster processes

 $graphs) \Rightarrow More flexible inputs$

A. Super-Kamiokande: B. Current State of PID Algorithms

5

A <u>B</u>oosted <u>D</u>ecision <u>T</u>ree (BDT) has been developed. Can we do better?

One would need to stack images to have the time information of an event

Small number of hits for the neutron capture on H (a GNN will not be

confused with useless information and therefore processed faster)

No direct relation between hits (one can add more complex information on

Merci Antoiliiiiⁱⁱⁱⁱⁱ
Merci Antoiliichitecture

Architecture example

-
-
- ▶ 2 layers of DynamicEdgeConv [1801.07829, Wang et al., 2019]: ▶ Connects closest (euclidian distance) nodes in feature space by edges ▶ Edge features: Information about x_i (node feature *i*) and $x_i - x_i$ (relative difference to nearest neighbours)

A. Super-Kamiokande: B. Current State of PID Algorithms

6

Setting the Low energy

2 My work

A. Key Challenges in PID

B. Adapting BDT Algorithm for High Energy

- Neutrinos
- Neutrinos

C. Adapting GNN Algorithm for High Energy

- e/mu particle identification
- e/gamma particle identification
- e/pi0 particle identification
- mu/pi+ particle identification
- Multiple ring fit

Identification of particles in levels of difficulties

2 A. Key Challenges in PID B. Adapting BDT

C. Adapting GNN

Muon (Sharp) Electron (Blurry)

My work **Distribution of Discriminating Variables**

I. DISTRIBUTION OF DISCRIMINANT VARIABLES

SEE IF THERE ARE ANY RELEVANT THID VARIABLES T O

My work **Distribution of Discriminating Variables Charge Profile**

My work **Distribution of Discriminating Variables Charge Profile**

Explication pour le mu :

- Broadening within the ring (see diagram) ٠ Hence, the peak shift is not exactly at 42°
- And thus, as the standard deviation is ٠ lower (compared to electron and gamma which are also broadened externally), the maximum value is higher.
- Charge per unit angle is higher for ٠ muons, dominating over the total charge since they are less scattered.

My work **Distribution of Discriminating Variables**

C. Adapting GNN

Charge Profile

My work Distribution of Discriminating Variables **Charge Profile**

My work **Distribution of Discriminating Variables Explanations t-TOF**

C. Adapting GNN

Calculation of t-tof as a function of particle progress x:

My work **Distribution of Discriminating Variables Explanations t-TOF**

My work Distribution of Discriminating Variables **Explications t-TOF**

My work **Distribution of Discriminating Variables**

C. Adapting GNN

My work **Distribution of Discriminating Variables**

Explication pour le mu: Shifted peak:

. Muon is about 2 times more energetic than each particle of the positron-electron pair, it will pass through the Cherenkov threshold later, so statistically, there is a higher chance that theta will be small, hence $x = x_{\text{tot}}$, resulting in negative t-tof. This shifts the peak.

Decreasing part:

- This corresponds to Cherenkov photons produced by the parent particle moving in its direction, after creation at the vertex.
- We see that Q decreases more quickly than for electrons. This is because muon passes through the Cherenkov threshold more rapidly than electrons.

Increasing part:

• Very steep, so few Cherenkov photons coming from charged particles whose trajectory is positively deviated.

C. Adapting GNN

My work **Distribution of Discriminating Variables**

Explication pour l'électron : Shifted peak:

• Same as for muon.

Decreasing part:

- This corresponds to Cherenkov photons produced by the parent particle moving in its direction, after creation at the vertex.
- We see that Q decreases more quickly than for electron. This is due to the fact that muon passes through the Cherenkov threshold more rapidly than electron.

Increasing and decreasing parts:

• Increasing part slightly less steep than for muon. Electromagnetic cascade populates both sides of the peak, so the decreasing part is also less steep.

C. Adapting GNN

My work Distribution of Discriminating Variables

Explication pour gamma:

- . Overall: distribution is much broader than for e-.
- Maximum charge per PMT: For the $e+/-$ pair, 2 times smaller than for e- and mu-. The maximum is found at t=tof. This seems consistent with the fact that when e+ and e- are very close to the vertex, the two generated photons are more likely to hit the same PMT. (The maximum charge per PMT does not allow for differentiation between e- and mubecause at the vertex there is no electromagnetic cascade effect, whereas, for the pair production, there is already a deviation in trajectory.)
- . In the increasing part, the same as for the electron.
- Wider distribution: for t-tof on either side of the peak, less steep rise and fall due to electromagnetic shower effect AND ADDITIONALLY angle between e+/e- direction.
- Conclusion: the difference between e- and $e+/-$ appears more pronounced concerning the charge profile because t-TOF is a variable equivalent to a length, unlike theta, thus we can better see the energy difference of the pair production.

C. Adapting GNN

My work **Distribution of Discriminating Variables**

Reconstruction - Charge Profile

C. Adapting GNN

Without randomization:

- . Mu is less scattered, so lower average std.
- Event by event: the std of gamma is larger than that of e because event by event gamma is more random due to pair production.
- The std of mu is generally larger because it fits Gaussian, and mu is skewed.

With random:

• Difficult to distinguish particles from each other... maybe try another fit?

My work **Distribution of Discriminating Variables**

ndusion

• For mu/e-: +: The difference in charge profile and the total charge is clearly visible. -: The difference is not very noticeable on the tof graphs.

For e-/gamma: \bullet the A clear difference is seen when studying the tof according to the charge per PMT. -: The difference is not very noticeable on the charge profile; it would be

necessary to investigate the angle of the pair production (i.e., perform exact calculations, but this is in progress...).

My work **Distribution of Discriminating Variables**

II. STUDY OF THE CORRELATION BETWEEN DISCRIMINANT VARIABLES

C'EST DES TH2D ICI DU COUP

My work **Distribution of Discriminating Variables**

C. Adapting GNN

bojween-variables

My work **Distribution of Discriminating Variables** Q(Theta, t-tof)

∴ Mu ٠

shower, resulting in a very clear break.

Charge peak at t-tof<0, but near 40°.

Electron

The charge is more scattered at a given t-tof, due to electromagnetic cascade. The peak at t-tof<0, near 42°. Distributed more or less uniformly around 42°, with a preferential direction.

More concentrated near 42°, few events at small t-tof after 42°. No scattering effect or electromagnetic

My work **Distribution of Discriminating Variables** Q(Theta, t-tof)

Gamma

Significantly weaker charge. At t-tof=0, highly scattered, and has no real preferential direction.

Electron

The charge is more scattered at a given t-tof, due to electromagnetic cascade. The peak at t-tof<0, near 42°. Distributed more or less uniformly around 42°, with a preferential direction.

My work **Distribution of Discriminating Variables Ratio Q(Theta, t-tof)**

My work **Distribution of Discriminating Variables** Ratio Q(Theta, t-tof)

My work **Distribution of Discriminating Variables Ratio Q(Theta, t-tof)**

Gamma

C. Adapting GNN

Mυ

My work **Distribution of Discriminating Variables**

III. STUDY OF THE IMPACT OF DETECTOR EFFECTS

RECONSTRUCTION EFFECT, WATER ABSORPTION EFFECT, AND RAYLEIGH SCATTERING EFFECT

My work Distribution of Discriminating Variables

2 A. Key Challenges in PID (B. Adapting BDT)

My work **Distribution of Discriminating Variables**

Q(theta) - Reconstruction Error

We study the influence of the reconstruction error on the charge profile with an error of:

- 20 cm on the reconstruction of the vertex position
- 2.5 degrees in the direction

Expected effect

We would expect the overall amplitude to be attenuated because changing the initial vertex position results in a more scattered distribution, thus decreased amplitude, increased standard deviation, and unchanged integral.

Observation

The charge profile, which takes this error into account, has a lower amplitude, but the mean is preserved: we can still distinguish between a mu and an e. The distinction between e and gamma is more difficult to make after randomization.

My work **Distribution of Discriminating Variables**

Q(TOF) - Reconstruction Error

 $\times 10^3$ 600 We study the influence of the reconstruction error on the charge 500 profile with an error of: • 20 cm on the reconstruction of the vertex position 400 • 2.5 degrees in the direction 300 _O **Expected effect** We would expect the overall 200 amplitude to be attenuated, with less effect on long-time photons because the further away from the vertex, the 100 less the initial position has an impact. 0 -10 **Observation** Amplitude attenuation on Q. No shift of the mean (ratio plot). We can still distinguish between gamma and e.

My work Distribution of Discriminating Variables

rption effects

2 A. Key Challenges in PID (B. Adapting BDT)

My work **Distribution of Discriminating Variables**

Q(theta) - Water absorption

Study of the influence of water absorption:

The characteristic length of 1.3 minus 1 sigma $= 0.07$.

The effect of water absorption

We would expect a major influence/ decrease on long-time photons populating the outer ring because those that pass through more material have a higher absorption probability. And less impact on gamma, more on mu and e.

Observations

No major changes, a small shift for the muon, because less in the outer ring.

My work **Distribution of Discriminating Variables** $Q(TOF) - Water$ absorption

Study of the influence of water absorption:

The characteristic length of 1.3 minus 1 sigma = 0.07 .

The effect of water absorption

We expect less population in the long-time, i.e., outer ring for mu and e (a little for gamma), and internal/ external for e and gamma due to the shower.

Observations

Weaker overall impact. Still, presence of a shift for mu and e and a slight shift to the left of gamma, so the distinction is possible between e and gamma.

My work Distribution of Discriminating Variables

My work Distribution of Discriminating Variables Q(theta) - Rayleigh scattering

My work Distribution of Discriminating Variables Q(TOF) - Rayleigh scattering

My work Distribution of Discriminating Variables

C. Adapting GNN

disfributions from

My work **Distribution of Discriminating Variables Cut for Charge Profile**

My work **Distribution of Discriminating Variables**

Cut for t-TOF Profile

My work **Hyper Parameters optimization**

My work **Hyper Parameters optimization**

My work **Hyper Parameters optimization**

C. Adapting GNN

Hyper Parameters optimization

51

C. Adapting GNN

Hyper Parameters optimization

A. GNN : ideas for improvement B. BDT : ideas for improvement

What's next?

GNN

3 A. GNN future improvements B. BDT future improvements

BDT

- Finishing the Optimization the hyperparameters of the GNN • Finding better cuts and study the correlation between built distributions
- Quantify the efficiency and precision of the GNN classification mu/e then identify the physical parameters of the GNN by comparing the distributions
- Do the same for e/gamma separation
- Optimize the parameters of the GNN at variable energy.
- Parallelization of the GNN training

• Take GNN's output parameters as inputs for the BDT

