# GraphNeT: Graph neural networks for neutrino telescope event reconstruction

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# Index of Arguments

- •Neutrino Physics
- •Neutrino Detection
- •Neutrino Experiments
- •IceCube drilldown and data analysis
- •GrapNet neural networks

## Introduction to Neutrinos

#### Introduction

•Neutrinos are subatomic particles that play a crucial role in our understanding of the universe.

•Studying neutrinos can help us learn more about fundamental forces, particle physics, and astrophysics.

•This presentation will cover the basic properties of neutrinos, their detections and the experiment IceCube. Then a new neural-network-based methodology will be presented for the event classification tasks.

#### **Standard Model of Elementary Particles**



2

## Charged Current Interaction Lagrangians

quarks	
$\mathcal{L}_{\text{cc}} = \frac{g}{\sqrt{2}} \overline{(u \ c \ t)}_{\text{L}} \ \gamma^{\mu} U \begin{pmatrix} d \\ s \\ b \end{pmatrix}_{\text{L}} W_{\mu}^{+} + \text{h.c.}$	Cabibbo-Kobayashi-Maskawa Matrix



Maki-Nakagawa-Sakata-Pontecorvo **Matrix** 

## V-Matrix Parametrization

$$
O_{1}(\theta_{1}, \alpha_{1}, \beta_{1}, \gamma_{1}) = \begin{pmatrix} c_{1}e^{i\alpha_{1}} & s_{1}e^{-i\beta_{1}} & 0 \\ -s_{1}e^{i\beta_{1}} & c_{1}e^{-i\alpha_{1}} & 0 \\ 0 & 0 & e^{i\gamma_{1}} \end{pmatrix}
$$
  
\n
$$
O_{2}(\theta_{2}, \alpha_{2}, \beta_{2}, \gamma_{2}) = \begin{pmatrix} e^{i\gamma_{2}} & 0 & 0 \\ 0 & c_{2}e^{i\alpha_{2}} & s_{2}e^{-i\beta_{2}} \\ 0 & -s_{2}e^{i\beta_{2}} & c_{2}e^{-i\alpha_{2}} \end{pmatrix}
$$
  
\n
$$
O_{3}(\theta_{3}, \alpha_{3}, \beta_{3}, \gamma_{3}) = \begin{pmatrix} c_{3}e^{i\alpha_{3}} & 0 & s_{3}e^{-i\beta_{3}} \\ 0 & e^{i\gamma_{3}} & 0 \\ -s_{3}e^{i\beta_{3}} & 0 & c_{3}e^{-i\alpha_{3}} \end{pmatrix}
$$
  
\nwhere  $s_{i} \equiv \sin \theta_{i}$  and  $c_{i} \equiv \cos \theta_{i}$  (for  $i = 1, 2, 3$ )

$$
\frac{\text{Dirac Neutrinos}}{V = O_i O_j O_i \ (i \neq j)}
$$

$$
\begin{array}{c}\n\text{Majorana} \text{Neutrinos} \\
\boxed{V = O_i O_j O_k \ (i \neq j \neq k)}\n\end{array}
$$

## Dirac or Majorana Neutrinos?

If neutrinos are Dirac particles, the phases  $x$ ,  $y$  and  $z$  can be removed. Then the neutrino mixing matrix is

#### **Dirac neutrino mixing matrix**



If neutrinos are Majorana particles, left- and right-handed fields are correlated. Hence only a common phase of three left-handed fields can be redefined (e.g.,  $z = 0$ ). Then

$$
V = \left(\begin{array}{ccc} c_{12}c_{13} & s_{12}c_{13} & s_{13}e^{-i\delta}\\ -s_{12}c_{23}-c_{12}s_{23}s_{13}e^{i\delta} & c_{12}c_{23}-s_{12}s_{23}s_{13}e^{i\delta} & s_{23}c_{13}\\ s_{12}s_{23}-c_{12}c_{23}s_{13}e^{i\delta} & -c_{12}s_{23}-s_{12}c_{23}s_{13}e^{i\delta} & c_{23}c_{13} \end{array}\right) \left(\begin{array}{ccc} e^{i\rho} & 0 & 0\\ 0 & e^{i\sigma} & 0\\ 0 & 0 & 1 \end{array}\right)
$$

# Neutrino Oscillations

- Neutrino oscillation is a spontaneous periodic change from one neutrino flavor state to another.
- It is a specific quantum phenomenon and it occurs as a natural consequence of neutrino mixing.



## Neutrino Mass Hyerarchy



# 2-Flavor Oscillation

#### The oscillation probability for appearance y experiments:

$$
P(\nu_{\mu} \to \nu_{e}) = |\langle \nu_{e} | \nu_{\mu}(t) \rangle|^{2} = |(\cos \theta \langle \nu_{1}| + \sin \theta \langle \nu_{2}|) (-\sin \theta | \nu_{1} \rangle + \cos \theta e^{-i\Delta E t} | \nu_{2} \rangle)|^{2}
$$
  
=  $|\sin \theta \cos \theta (1 - e^{-i\Delta E t})|^{2} = 2 (\sin \theta \cos \theta)^{2} (1 - \cos \frac{\Delta m^{2} t}{2E})$   
=  $\sin^{2} 2\theta \sin^{2} \frac{\Delta m^{2} L}{4E}$ 

#### The conversion and survival probabilities in realistic units:

$$
P(\nu_{\mu} \to \nu_{e}) = \sin^{2} 2\theta \sin^{2} \frac{1.27 \Delta m^{2} L}{E}
$$

$$
P(\nu_{\mu} \to \nu_{\mu}) = 1 - \sin^{2} 2\theta \sin^{2} \frac{1.27 \Delta m^{2} L}{E}
$$

Due to the smallness of  $(1,3)$  mixing, both solar & atmospheric neutrino oscillations are roughly the 2-flavor oscillation.

 $\Delta m^2$  in unit of  $eV^2$ , L in unit of km, E in unit of GeV

# 1.27 ?



# 3-Flavor Oscillation

#### The final formula of 3-flavor oscillation probabilities with CP violation:

$$
P(\nu_{\alpha} \to \nu_{\beta}) = \delta_{\alpha\beta} - 4\sum_{i
$$

The 1<sup>st</sup> oscillating term: CP conserving; and the 2<sup>nd</sup> term: CP violating!

#### Neutrino Oscillation Probabilities



 $L$  [ $km$ ]

# Neutrino Oscillation Studies

#### $v_{\mu} \rightarrow v_{\tau}$  oscillations ( $\Delta m_{23}$ ,  $\theta_{23}$ ) Atmospheric: Super-K, Soudan-2, MACRO IceCube/Deepcore, ... LBL: K2K, MINOS, OPERA, T2K, NOvA, ...

 $v_e \rightarrow (v_u + v_t)$  oscillations ( $\Delta m_{12}$ ,  $\theta_{12}$ ) Solar: SNO, Super-K, Borexino, ... **Reactor: KamLAND** 

 $\theta_{13}$  experiments LBL: MINOS, T2K, NOvA, ... Reactor: Daya Bay, Reno, Double Chooz

#### **Status (before Neutrino 2016)**



K. Nakamura and S.T. Petcov, "14. Neutrino mass, mixing and oscillations"

# Index of Arguments

- •Neutrino Physics
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- •Neutrino Experiments
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#### How to Detect Neutrinos: Cerenkow Light

- Cerenkov radiation is emitted whenever a charged particle passes through a (dielectric) medium with velocity βc=v>c/n, where v is the velocity of the particle and n the refractive index of the medium.
- Incoming particle polarizes atoms in the medium which, in turn, become electric dipoles:



# Cerenkon Light

• The emission angle  $\theta_c$  can be qualitatively interpreted as a shock wave as happens for a boat or a supersonic plane.



- Threshold velocity:  $\beta_s = 1/n \rightarrow \theta_c \sim 0$
- Maximum angle:  $\theta_{\text{max}}$ = arccos(1/n) In water n = 1.33 and  $\theta$ c = 41°

The above equation are valid in L>> $\lambda$  where L is the length of the medium and  $\lambda$  is the wavelength of emitted light

#### Cerenkov Spectrum

Number of photons emitted per unit length and unit wavelength range. We observe that it decreases as λ increases.

$$
\frac{d^2N}{dx d\lambda} = \frac{2\pi z^2 \alpha}{\lambda^2} \left( 1 - \frac{1}{\beta^2 n^2} \right) = \frac{2\pi z^2 \alpha}{\lambda^2} \sin^2 \theta_C
$$
  

$$
\frac{d^2N}{dx d\lambda} \propto \frac{1}{\lambda^2} \quad \text{with} \quad \lambda = \frac{c}{v} = \frac{hc}{E} \quad \frac{d^2N}{dx dE} = const.
$$
  

$$
\alpha = \frac{2 \pi e^2}{hc} = \frac{1}{137} \frac{1}{4\pi \epsilon_0} = 1
$$

The number of photons emitted per unit of Length does not depend on energy.



# Cerenkon Light Energy Loss

• The energy lost by Cerenkov radiation increases with β. However even with β -> 1 it is very small, and generally smaller than Bethe Block energy loss:

$$
-\frac{dE}{dx} = z^2 \alpha \frac{\hbar}{c} \int \omega \left(1 - \frac{1}{\beta^2 n^2(\omega)}\right) d\omega
$$



# How much light?

Let us consider:

- a 1 cm thick radiator
- an angle  $\theta$ c = 30 $\circ$
- $a$  DE = 1 eV
- a particle of charge = 1

$$
\frac{dN}{dEdx} = \frac{z^2 \alpha}{\hbar c} \sin^2 \theta_c
$$
  
\n
$$
\implies N_{ph} = 370 \cdot \sin^2 \theta_c \cdot L \cdot \Delta E = 370 \times 0.25 = 92.5
$$

Considering, also, that the quantum efficiency of a photomultiplier is around 20% then:

$$
\Rightarrow N_{ph}=18
$$

# Index of Arguments

- •Neutrino Physics
- •Neutrino Detection
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## Cherenkov Neutrino Experiment Timeline

- DUMAND: Deep Underwater Muon and Neutrino Detector proposed by F. Reynes on 1975 (9 strings and 216 OMs, 4.6 Km deep of bottom level).
- BAIKAL Bezrukov, Domogatsky, Berezinsky, Zatsepin 1981 First site exploration and R&D, 1.3 Km depth, Baikal lake, from NT36 (36 OMs) to NT200 (200 OMs)
- NESTOR Pylos island, Greece 3.8 Km deep. Survey started in the 90's (12 floors with 14 OMs per floor, 168 in total)
- AMANDA South Pole 1996 started (19 strings and 677 OMs)
- ANTARES R&D started in 1997, off-shore of Toulon, 2.5 Km deep (12 strings, 75 OMs
- each string)
- NEMO R&D start in 1998 first site exploration in 2002, off-shore of Capopassero, 33.5 Km deep, phase 1 (4 floors 16 OMs), phase 2 (8 floors 32 OMs)
- IceCube construction 2005-2010, South Pole, 86 strings, 5160 OMs
- KM3NeT/ARCA construction started on Dec 2015, off-shore Capopassero, 2 building blocks with 115 strings per block and 18 Doms each string.

# KM3NeT/ARCA Overview



## ARCA Reconstructed Events (muon)



- The size of the spheres reflects the amount of light collected and their colour represents the arrival time of the light on the module: red is earlier than blue.
- The path of the muon particle that travels from above is reconstructed from this information and is made visible as the blue line.

## ARCA Simulated Events



# Index of Arguments

- •Neutrino Physics
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- •Neutrino Experiments
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#### IceCube Overview



## IceCube Overview

- The cables of all 86 strings run together in the IceCube Lab
- On-line processing and filtering
- Detector uptime of 99.8% (!)
- Data transfer to North:
	- High priority data (e.g. alerts) can be sent 24/7 over IRIDIUM connections (very low bandwidth), Starlink in testing phase
	- Usually, a couple hours per day satellites with higher bandwidth are in reach, can transfer up to  $~100$ GB/day
- Rest of data is literally "shipped" out on disk



## Events in IceCube

- Every DOM gets around ~500-800 hits per second, mainly from dark noise
- Hits from physics events are  $\sim$ 1 order of magnitude fewer
- Most of this is suppressed by trigger conditions
- Per year, we read out roughly:
	- $\cdot$  10<sup>10</sup> events caused by atmospheric muons
	- 10<sup>9</sup> events caused by noise
	- 100.000 events from atmospheric neutrinos
- "A handful of very high energy events likely to be of astrophysical origin"
- Special triggers exist for example looking for supernovae, they monitor the overall hit rate, where a correlated increase could indicate a nearby supernova

#### IceCube Event Reconstruction



# Index of Arguments

- •Neutrino Physics
- •Neutrino Detection
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## CNNs in Neutrino Telescopes

#### • Transform the raw data into 3d images • A convolutional neural network



- based cascade reconstruction for the IceCube Neutrino **Observatory**
- https://doi.org/10.1088/1748- 0221/16/07/P07041

→**One big limitation: transformation onto rigid 3d "image"**  → **Especially for DeepCore**

# Graph Neural Networks





But, the data / detector geometry may deviate from this regularity

- **CNN:** fixed, rigid correlation structure: **GGN:** Representation of data as a mathematical graph:
	- **Node** features can be exactly the same as pixels in CNNs
	- **Edges** representing the pair-wise relation between nodes (e.g. spatial distance)

- **G**raph **N**eural **N**etworks (**GNN**s):
	- generalization of NNs to arbitrary geometries
	- can define "graph convolutions" similar to CNNs





[graphnet-team/graphnet: Graph](https://github.com/graphnet-team/graphnet/tree/main) neural [networks for neutrino telescope](https://github.com/graphnet-team/graphnet/tree/main) event reconstruction (github.com)

# GraphNet Workflow



#### High-level overview of a typical workflow using GraphNeT:

1.*graphnet.data* enables converting domain-specific data to industry-standard, intermediate file formats and reading this data;

2.*graphnet.models* allows for configuring and building complex GNN models using simple, physicsoriented components;

- 3. *graphnet.training* manages model training and experiment logging;
- 4. *graphnet.deployment* allows for using trained models for inference in domain-specific reconstruction chains.

# DynEdge: the Heart of GraphNet

**Input Graph** 



# DynEdge Train & Test

- Low energy neutrino dataset from 1GeV to 1000GeV simulated with GENIE <https://doi.org/10.1016/j.nima.2009.12.009>
- Training on the simulated dataser with separation of "track-like" events from "cascade-like" ones.
- IceCube detector has 5160 sensors at fixed, known locations
- Per event each sensor receives zero many "pulses", i.e. variable-length time series
- Comparison of results with RETRO algorithm <https://doi.org/10.1140/epjc/s10052-022-10721-2>



#### IceCube Event Reconstruction Sample



#### DYNEDGE classifier vs Boosted Decision Tree performance

- 18% increase in  $\nu/\mu$  classification task
- 6% increase in AUC score in  $T/C$  classification task

# The End

