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Physics potential of the IceCube Upgrade for atmospheric neutrino oscillations

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The IceCube Upgrade is an extension of the existing IceCube Neutrino Observatory and will be deployed in the 2025–2026 austral summer. It will significantly improve the sensitivity of the detector to atmospheric neutrino oscillations. The existing 86-string IceCube array contains a dense in-fill known as DeepCore which is optimized to measure neutrinos with energies down to a few GeV. The IceCube Upgrade will consist of seven new densely instrumented strings placed within the DeepCore volume to further enhance the performance in the GeV energy range. The additional strings will feature new optical modules, each containing multiple photomultiplier tubes (PMTs), in contrast to the existing modules that each contain a single PMT. This will more than triple the number of PMT channels with respect to the current IceCube configuration, allowing for improved detection efficiency and reconstruction performance at GeV energies. We describe necessary updates to simulation, event selection, and reconstruction to accommodate the higher data rates observed by the upgraded detector and the addition of multi-PMT modules. We determine the expected sensitivity of the IceCube Upgrade to the atmospheric neutrino oscillation parameters $\sin^2\theta_{23}$ and Δm_{32}^2 , the appearance of tau neutrinos and the neutrino mass ordering. The IceCube Upgrade will provide

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neutrino oscillation measurements that are of similar precision to those from accelerator experiments, while providing complementarity by probing higher energies and longer baselines, and with different sources of systematic uncertainties.

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I. INTRODUCTION

Studying the oscillations of atmospheric neutrinos provides a unique window into the nature of these elusive particles. Since neutrinos arrive at the detector from all directions, many different oscillation baselines are probed simultaneously ranging from tens of kilometers up to the diameter of the Earth. This can be used to make precise measurements of the atmospheric neutrino oscillation parameters, such as the previous measurements in [1,2], as described by the Pontecorvo-Maki-Nakagawa-Sakata (PMNS) mixing matrix [3,4]. Figure 1 shows the probability of an atmospheric muon neutrino oscillating into each flavor as a function of energy and zenith angle, where zenith angle is a proxy for distance traveled. The primary signature is the disappearance of muon neutrinos and the appearance of tau neutrinos. There is negligible production of tau neutrinos in the atmosphere, and thus detected tau neutrinos in this energy region arise from oscillations. The primary physics parameters Δm_{32}^2 and θ_{23} control the location and amplitude of the peaks, respectively. The distortions visible below $\cos(\theta_{zen}) < -0.8$ corresponds to the region where neutrinos pass through the core of the Earth, which has much higher density. These matter effects are a subdominant component that occur in all flavors primarily below 10 GeV, however they are most noticeable in the electron neutrino channel since the other channels are dominated by the $\nu_\mu \rightarrow \nu_\tau$ oscillation. In addition, the different effects of matter on neutrinos and antineutrinos traversing the Earth

can be used to distinguish the ordering of the neutrino mass eigenstates [5]. Therefore, the neutrinos in this region below 10 GeV will be crucial for measurements of the neutrino mass ordering.

Atmospheric neutrinos are generated in cosmic-ray interactions with nucleons in the Earth's atmosphere. When interacting in optically clear media, such as water or ice, the interactions of these GeV neutrinos produce charged secondary particles that emit Cherenkov radiation. The emitted photons can be seen by ice- or water-Cherenkov detectors such as the IceCube Neutrino Observatory [6] which is located at the geographic South Pole. IceCube currently consists of 5160 photosensors, known as digital optical modules (DOMs), spread across 86 long cables, known as strings, each deployed in a separate borehole in the ice.

The primary IceCube array is spread throughout a cubic-kilometer volume and was designed to observe high-energy neutrino interactions at the TeV to PeV energy scale. These strings are arranged in a hexagonal grid. The existing low-energy extension, known as IceCube DeepCore [7], consists of eight strings with more dense optical module spacing and high quantum efficiency photomultiplier tubes (PMTs). The DeepCore modules are located in the deepest and clearest portion of the ice. The optical module spacing in DeepCore allows for the detection of neutrinos between roughly 5 to 500 GeV at a rate of $\mathcal{O}(\text{mHz})$ at analysis level. DeepCore has already collected about a decade of atmospheric neutrino data, with recent neutrino oscillation measurements reported in [1,8].

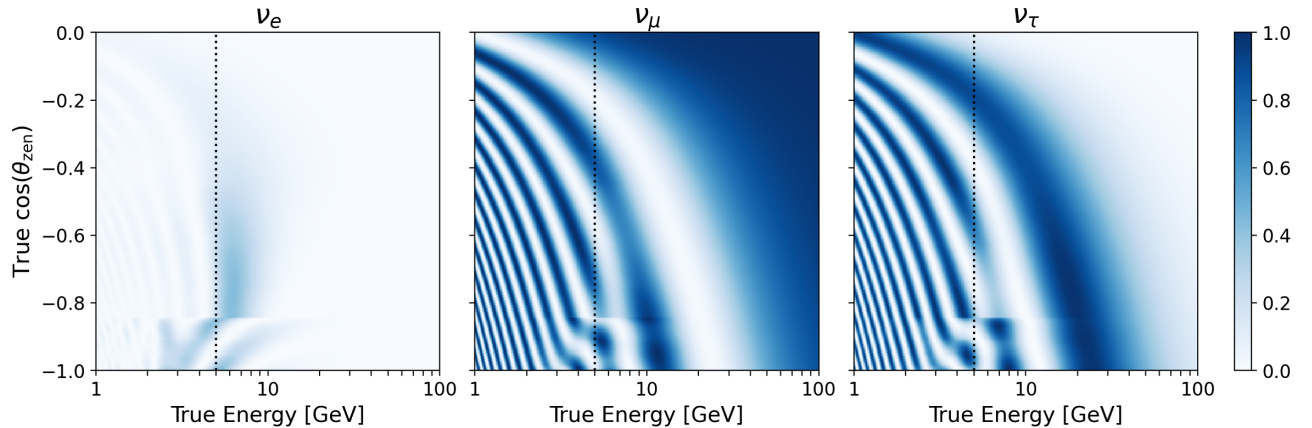


FIG. 1. Probability of a muon neutrino oscillating to each flavor (left: electron neutrino, center: muon neutrino, right: tau neutrino) as a function of true energy and arrival direction. $\cos(\theta_{zen})$ is a proxy for the distance that the neutrino traveled. The IceCube Upgrade will make the region below 5 GeV, indicated by the dotted line, much more accessible to analyses.

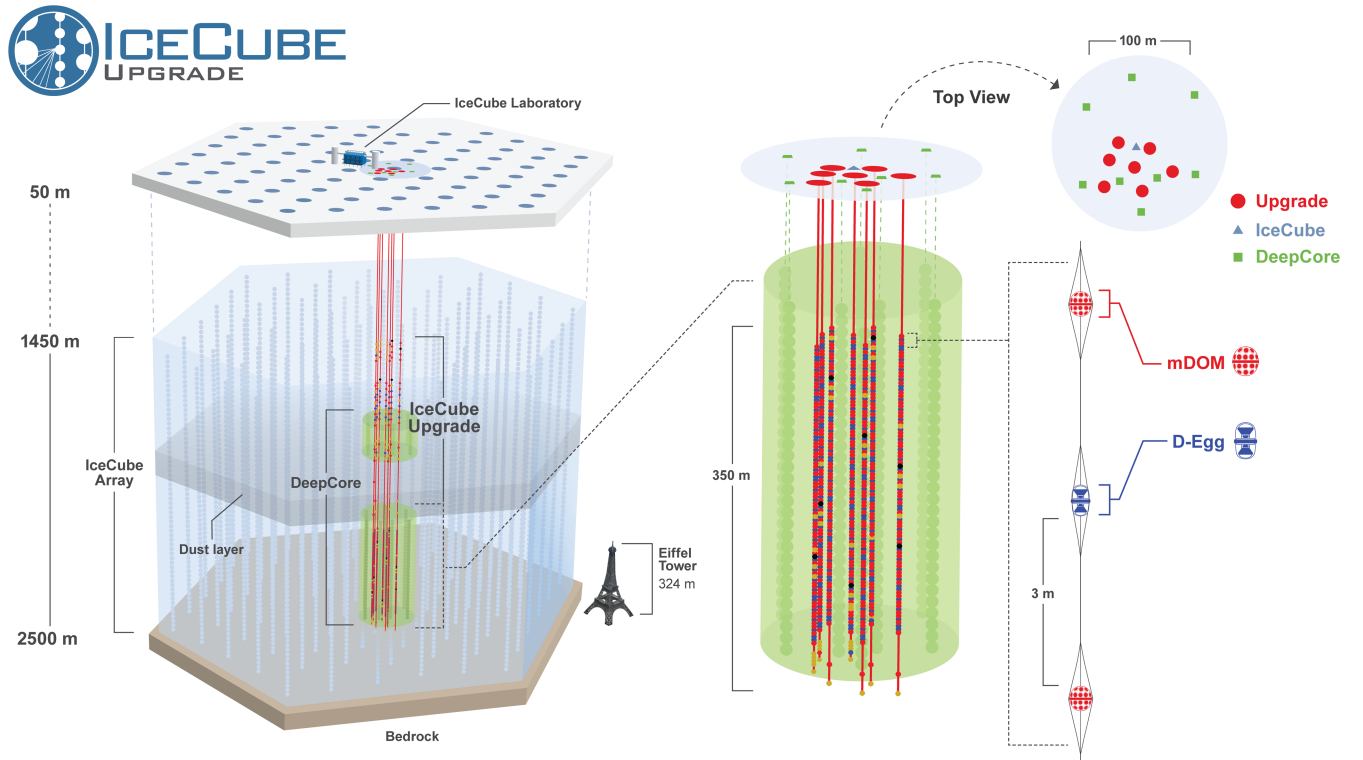


FIG. 2. The IceCube Upgrade will consist of 7 new strings, to be deployed in 2025–2026, at the locations shown in the top view in red. The current IceCube array optimized for TeV neutrinos is shown in blue, while the existing low-energy extension known as DeepCore is shown in green. The side view shows the distribution of module types on the new strings. The main optical modules listed in Table I are shown. The modules shown in yellow and black correspond to the R&D optical modules described at the end of Sec. II B and the calibration modules described in Sec. II C. Most of the modules in the IceCube Upgrade are located in the deepest part of the ice, but there are some modules in the upper portion of IceCube that will primarily be used for calibration.

To lower the energy threshold and increase the rate of detected neutrinos, seven additional strings will be deployed within the DeepCore fiducial volume in the 2025–2026 austral summer. The installation of these additional strings, known as the IceCube Upgrade, will increase the total number of strings in IceCube from 86 to 93. This fully funded IceCube Upgrade extension is distinct from the previously proposed “PINGU” extension (precision IceCube next generation upgrade) [9] which is no longer being pursued.

The additional strings will be equipped with new optical modules, each hosting multiple PMTs, as well as several new calibration devices. After installation, the number of PMT channels will be more than three times higher than the current detector configuration. Additionally, the total photocathode area in the DeepCore fiducial volume will increase by a factor of about three. The effective photosensitivity increase will be even higher than this due to improvements in module design that increase the acceptance for shorter wavelengths, discussed in more detail in Sec. II B. These improvements together with closer module spacing will greatly enhance IceCube’s capabilities in the GeV energy regime.

This paper discusses the IceCube Upgrade detector and simulation (Sec. II), new tools developed for handling IceCube Upgrade events (Secs. III and IV), and shows the expected enhancement in sensitivity for several benchmark oscillation analyses by comparing a scenario with and without the additional strings (Secs. V and VI).

II. THE ICECUBE UPGRADE DETECTOR

A. Geometry

The IceCube Upgrade will consist of seven new strings deployed within the DeepCore fiducial volume, as shown in Fig. 2. The existing spacing between strings in the DeepCore fiducial volume is about 40–70 m. With the IceCube Upgrade, the typical inter-string spacing in this region will be reduced to about 20–30 m. In addition, the new modules will be more closely spaced in depth. On the IceCube strings, the modules are spaced 17 m apart vertically spanning depths of about 1450 to 2450 m below the surface. The modules on the DeepCore strings are spaced 7 m apart and are concentrated in the deepest part of the ice from 2100 to 2450 m, where the ice has the best optical properties. The modules on the upgrade strings will

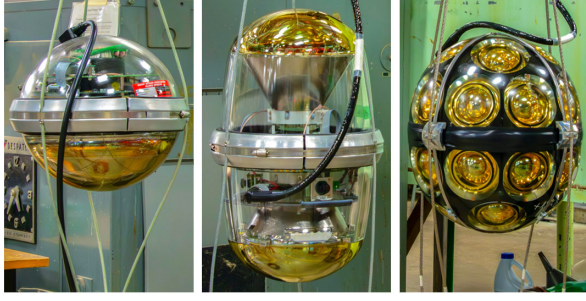


FIG. 3. Photographs of the modules: IceCube DOM (left), D-Egg (middle), and mDOM (right).

be spaced 3 m apart, with most of the modules concentrated between 2160 m and 2430 m.

To improve our understanding of the optical properties of the ice, there will also be modules between 1300 m and 1900 m as well as below 2450 m which will be used for calibration purposes. The existing IceCube plus DeepCore configuration is referred to as “IC86.” The configuration including the new IceCube Upgrade strings is referred to as “IC93.”

B. Optical modules

The IceCube Upgrade strings will be equipped with new types of optical modules. Two different sensor types form the majority of the approx. 700 new optical modules that will be deployed. There will be 280 “D-Eggs” (Dual optical sensors in an Ellipsoid Glass for Gen2) [10] each containing two 8-inch R5912-100-70 PMTs, and 400 “mDOMs” (multi-PMT Digital Optical Module) [11–13] each containing 24 3-inch R15458-02 PMTs [14]. The distribution of the new modules across the detector is shown in Fig. 2. Photographs of the modules are shown in Fig. 3.

Both modules have several advantages over the existing IceCube and DeepCore modules with only one 10-inch PMT each. The multiple-PMT configuration provides for a more uniform solid angle coverage, increases the photocathode area per module, and offers an intrinsic directional resolution. D-Egg and mDOM PMTs were selected to have low dark noise rates, high quantum efficiency, and excellent charge/timing resolution. Key parameters of these modules are summarized in Table I.

D-Eggs and mDOMs have fast, low-power electronics that digitize PMT analog signals at 240 MHz and 120 MHz, respectively, to achieve nanosecond precision. The readout

system has no dead time using internal 2 GB DDR3 RAM for continuous data acquisition. The high noise rates from multiple PMTs, however, would saturate the communication bandwidth of each string’s cable. Microcontrollers inside the modules partially mitigate this effect through lossless compression, feature extraction of single photon events, and in-module local coincidence.

Both module types incorporate several calibration devices. Each module has LEDs that produce nanosecond pulses pointing at various orientations which will be used for PMT gain calibration and ice property inspection using neighboring modules [15]. They are also equipped with fisheye cameras and illumination LEDs to monitor the refrozen ice surrounding the strings [16], as well as environmental sensors to monitor pressure, temperature, magnetic field, and acceleration.

Each module’s components are housed in a pressure-tolerant glass vessel. The PMTs are optically coupled to the vessels using silicone gel for increased photon transparency and mechanical support. The vessels and the gels were selected based on their UV transparency, greatly enhancing the transmittance at shorter wavelengths compared to the existing modules, as well as their low radioactivity. For the D-Egg, the overall high UV transparency, together with the increased photocathode area and the higher quantum efficiency of the PMTs, leads to the Cherenkov-averaged photon sensitivity about 2.8 times higher than the existing IceCube DOMs [10]. The D-Egg and mDOM glass vessels have diameters of 300 mm and 356 mm without the metal harnesses for installation, respectively, comparable to the existing 330 mm DOMs.

In addition to D-Eggs and mDOMs, some other types of optical modules will be deployed for R&D purposes. These modules include pDOMs (DeepCore-like single PMT DOMs), WOMs (wavelength-shifting optical modules), LOMs (long optical modules), FOMs (fiber optic module). pDOMs each contain one high quantum efficiency PMT and a muon-tagging assembly. WOMs [17] use wavelength-shifting paint in a transparent tube to increase the photon detection rate at short wavelengths. LOMs [18,19] are a hybrid between the D-Egg and mDOM each containing 16 or 18 4-inch PMTs with a narrower diameter than the existing IceCube modules. The FOM uses fluorescent optical fibers to guide photons to a PMT. The R&D modules are not included in the sensitivity calculations that follow.

TABLE I. Key parameters of the two main upgrade detectors D-Egg and mDOM compared with the existing IceCube DOM.

Name	No. of modules	Diameter (mm)	PMT size (inch)	No. of PMTs	No. of flasher LEDs	No. of cameras
D-Egg	280	300	8	2	12	3
mDOM	400	356	3	24	10	3
IceCube DOM	5160	330	10	1	12	0

C. Calibration modules

The IceCube Upgrade strings will also feature a suite of about 50 new calibration devices aimed at improving the characterization of our detection medium, the glacial Antarctic ice. The precision optical calibration modules, or “POCAMs” [20,21], are isotropic light sources providing nanosecond light pulses that can be used for optical calibration. “Pencil beam” modules [22] produce a laserlike beam that can be pointed in almost any direction, allowing for more precise directional and anisotropy calibrations. Acoustic modules [23] will be used for the precise calibration of the positions of the optical sensors, and camera systems [16] will observe the optical properties of the ice. In addition, seismometers [24] allow for a broad spectrum of Earth science studies.

III. SIMULATION

A. Particle interaction and propagation

The tools used to produce Monte Carlo (MC) simulation for the IceCube Upgrade are similar to those used for previous IceCube DeepCore measurements [8]. The simulation chain has been improved since the previous IceCube Upgrade public data release [25]. It now includes a more realistic detector response and improved reconstruction and processing.

The simulation chain consists of three steps: primary particle interaction, photon (and other particle) propagation, and detector response. We simulate three different types of events: signal neutrino events and backgrounds from atmospheric muon events and pure noise events. The latter arise from random coincidences of PMT noise.

Neutrino events are generated using GENIE version 2.12.8 [26,27] following an $E^{-2.5}$ energy spectrum. Future iterations of this work will use a more modern generator and tune, and the impact of additional relevant cross section systematic uncertainties will be tested. They are produced throughout a cylindrical volume surrounding the DeepCore region, which also includes the new upgrade strings. We simulate all three neutrino flavors, neutrinos and antineutrinos, charged current (CC), and neutral current (NC) interactions. Neutrino events are weighted to a nominal atmospheric neutrino flux from Honda *et al.* [28] plus a three-flavor oscillation probability including matter effects.

Atmospheric muons are simulated using MuonGun (based on [29]) which produces muons on the surface of a cylinder surrounding IceCube and then weighted to CORSIKA [30]. Computation time is reduced by requiring that the muon reaches an inner cylinder surrounding the DeepCore/upgrade region. The energy and zenith angle (the angle between the muon direction and the Earth’s axis) of the generated muons are biased to prioritize muons that are most likely to survive the later stages of the event selection, following the procedure introduced in [31]. The muon

events are reweighted accordingly to avoid bias in the final event distributions.

After the primary particle interaction, muons from ν_μ CC interactions and atmospheric muons are propagated using PROPOSAL [32]. All other particle types are propagated and decayed using Geant4 [33]. The charged particles from these interactions with sufficiently high energy will emit Cherenkov radiation. The resulting Cherenkov photons are then propagated through the ice using CLSim [34] according to the ice model described in [35].

Muons travel about 4.5 m per GeV in ice and therefore muons detected by the IceCube Upgrade can travel several tens of meters. In contrast, other leptons and hadrons travel much shorter distances. As a result, the distribution of light in the detector looks different depending on whether a muon was present or not. Consequently, ν_μ CC interactions as well as atmospheric muons produce an elongated event topology which we refer to as “tracks,” while all other interaction types will produce rather spherical events, denoted “cascades.” Neutrinos and antineutrinos are not distinguished in the detector on an event-by-event basis; only the sum of both contributions is observed.

B. Detector response

Once a simulated photon reaches the surface of a new module, the following detector response is applied. First, the photon detection probability is determined based on the module angular acceptance [36] and quantum efficiency. If the photon is detected, it is converted into a photoelectron (PE). In addition to these signal PEs, noise PEs from PMT thermal dark noise and radioactive decays in the glass pressure housing and PMT glass are added. PEs that are closer in time than 0.2 ns are merged since these would not be resolved separately by readout electronics. Then, the PEs are given to a PMT response simulation which outputs “pulses” that have a charge and a time value assigned. The charge is randomly drawn from a Gaussian + exponential PMT charge distribution [6], while the time is the first PE time smeared by the PMT transit time spread. Pulses with a charge of less than 0.25 of the average single PE charge are removed because such pulses would not be distinguishable from noise in the electronics. Finally, a Gaussian smearing with a width of 1 ns is applied to the time of each pulse to account for the readout timing resolution. A second Gaussian smearing with a width of 1 ns is applied to all pulses in a module simultaneously to emulate the synchronization timing resolution.

In contrast to this simplified detector response, for the existing IceCube DOMs the actual waveform is simulated and unfolded to extract pulse information (see Sec. 4 in [37]). The detector response of the new modules will be updated to reach the same level of detail in the future.

Random pure noise triggers are simulated as well. Since they do not involve a primary particle nor a photon

propagation step, pure noise events only go through the detector response step and only involve noise PEs.

IV. SIMULATED DATA PIPELINE

A. Event trigger

Once the pulses are recorded, a trigger is used to look for spatial and temporal coincidences to determine whether an event occurred, and if so, which pulses belong to it. Two different triggers are used and events are kept if either one or both of them fired. One is the DeepCore simple majority trigger “SMT3” described in [7] which only uses existing IceCube and DeepCore strings, while the second uses the new upgrade strings and is described in the following.

Pulses that are near each other in space and time are used to establish “local coincidence” (LC), indicating that they could be causally connected. Different coincidence criteria are used for checking multiple PMTs on the same optical module and for multiple optical modules on the same string. For the first case, a pair of pulses are defined as coincident if they are on same module but on different PMTs and their time difference is less than 10 ns. Two pulses are also coincident if the modules are close to each other on the string, defined as ± 8 modules (~ 24 m), and if their time difference is less than 250 ns. This corresponds approximately to the speed of light in ice. The pulses with “local coincidence” are then used to check for triggers. If 8 or more LC pulses occur within a 1750 ns window, the trigger fires. All pulses from 4 μ s before to 6 μ s after the trigger condition is met are read out. The dominant event types triggering the detector are atmospheric muons $O(100$ Hz) and pure noise events $O(10$ Hz). Further levels of processing are necessary to reduce these backgrounds, and are described in Sec. IV D.

B. Removal of noise pulses

In the past, IceCube has been relying on cuts on the spatial and temporal distance between pulses to remove noise pulses from events. However, due to the increased number of PMTs in the new modules and higher than anticipated level of radioactivity in the mDOM PMT glass, these cut-based methods have been insufficient for noise removal. To address this, we have developed a new “noise cleaning” algorithm based on graph convolutional neural networks (GNNs). GNNs work on *graphs* and can be thought of as generalized convolutional neural networks. A graph is a collection of nodes and edges, where nodes typically represent data points and edges denote a relationship between the nodes, defining how information may spread during convolutions [38].

We chose to represent IceCube events as point cloud graphs, where nodes represent each PMT pulse recorded in an event and edges are drawn to the 8 nearest neighbors of each node based on the Euclidean distance between PMTs. The features associated with each node are shown in

TABLE II. Input features used for training pulse cleaning and reconstruction methods.

Feature	Description	Dimension
P_{xyz}	PMT position	$\mathbb{R}^3$
D_{xyz}	PMT orientation unit vector	$\mathbb{R}^3$
t	Estimated arrival time at PMT	$\mathbb{R}$
q	Estimated PE charge	$\mathbb{R}$
rde	Relative DOM efficiency	$\mathbb{R}$
A	Area of PMT	$\mathbb{R}$
idx	String, DOM and PMT number	$\mathbb{Z}^3$
$type$	DOM type	$\mathbb{Z}$

Table II. Using this graph representation, we then perform binary classification to classify each node (pulse) in each event as either signal or noise. We train a modified version of DynEdge [39] from the open-source deep learning library for neutrino telescopes GraphNeT [40]. This new noise cleaning method produces a prediction between 0 and 1 for each pulse in an event, where 0 indicates a pulse is likely noise and 1 indicates a pulse is likely signal. The method is compatible with events containing any number of pulses.

Because the output score is continuous, the amount of cleaning can be easily tuned by changing the threshold. Once trained on an independent sample of close to 4 million neutrino and muon events in IceCube Upgrade, we find that, at a threshold of 0.7, the GNN on average reduces the amount of noise by roughly a factor 10 with only minor loss ($\sim 5\%$) in signal hits. The contamination of noise for pulses that pass the cleaning is 6.8% using the new GNN method, whereas for the traditional noise cleaning method applied to the upgrade it was 70%. The previous method looks for hits that could be causally connected and is described in [8]. The retained signal vs optical module noise rate is shown for the GNN-based cleaning method and compared to the traditional cleaning method in Fig. 4. As seen in Fig. 4, the GNN cleaning method retains a higher percentage of the signal than the traditional method, and the remaining noise rate depends on the module type. The pulses that remain after cleaning are referred to as the “cleaned pulse series.”

C. Event reconstruction

Four reconstruction and classification tasks are trained using different instances of DynEdge with a procedure similar to that outlined in [39]. In order to mitigate class imbalances during training, task-specific subsamples produced from the full simulation sample are used for training. These samples differ primarily in their composition of neutrino track (T) events, neutrino cascade (C) events, and muon (μ) background events, as seen in Table III. The *Reco* sample contains neutrino events with at least three pulses in

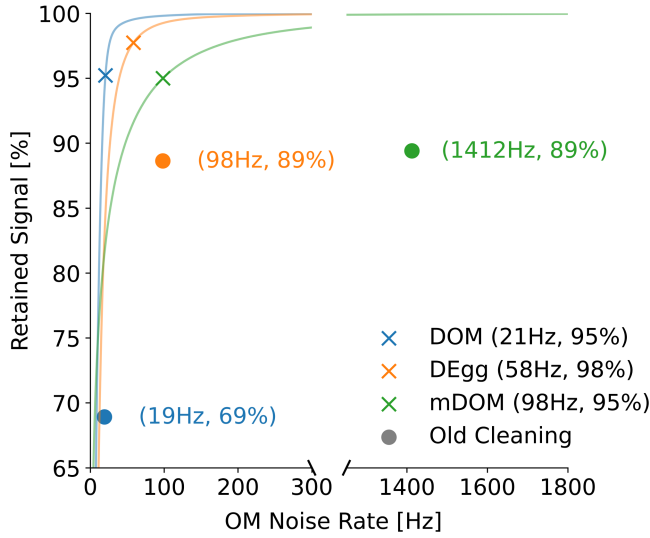


FIG. 4. Retained signal vs module noise rate. The retained signal is the true positive rate in percent, and the optical module noise rate is computed as the false positive rate multiplied by the total optical module noise rate. Each curve represents different optical module types. Crosses denote the classification threshold of 0.7 used for this analysis. The performance of the previous cleaning method for each module type is added as a circular dot in the respective color for reference.

the cleaned pulse series. The ν/μ sample was created by first subsampling the whole sample to obtain an equal number of neutrino and muon events and subsequently requiring three pulses in the cleaned pulse series. An identical procedure was used to construct the T/C sample, but ν_τ CC events have been omitted, as around 17% of these interactions may produce tracklike signatures, which may affect training [39,41].

In contrast to the noise cleaning task, where each pulse is assigned an output score, event classification and reconstruction are run on a per-event mode, where each model produces a single prediction for each event. The first model is trained to estimate neutrino energy and uses the *Reco* sample mentioned in Table III. Another model is trained to estimate the zenith angle of the incoming neutrino, along with an uncertainty estimation on the zenith

prediction. The zenith reconstruction network is also trained using the *Reco* sample.

The remaining tasks are classification. For the third event-level task an instance of DynEdge is trained to classify events as tracklike or cascadelike using the T/C sample. The fourth event-level task is classifying between muon and neutrino events, which is used in the event selection to reject background muons from the neutrino sample. This is trained on the ν/μ sample. To mitigate class imbalance, the ν/μ sample has a roughly equal number of muon and neutrino events in it. Due to the computational intensity of producing background muon simulation, we have a relatively low number of simulated muons, and therefore this sample is the smallest at just about 367,000 events.

Selected performance metrics and their energy dependence are shown in Fig. 5 for the zenith and energy reconstruction models, along with the energy dependence of the T/C classifier’s performance. Figure 5 compares the performance metrics between IC86 and IC93. However, because the threshold for detection by IC86 is higher than IC93, the IC93 sample (orange) includes low-energy events that were not detectable by IC86; these low-energy events tend to produce only a few pulses in the detector and are, therefore, hardest to reconstruct. In order to directly quantify the improvement due to the new strings on the exact same set of events, we identify a subset of events detectable by IC86 that have at least one pulse in IC93. The events in this subsample detectable by IC86 are then reconstructed with and without the information on the new strings, denoted “IC86 Subs. w/ upgrade” (green) and “IC86 Subs.” (blue) respectively. Comparing the green and blue thus represents the improvement on the exact same events directly due to the inclusion of the new strings. By contrast, the orange curve contains all events in the sample and therefore additional effects such differences in the sample are introduced. Additional details about these reconstruction comparisons can be found in Appendix B.

D. Event selection

At trigger level, the events are dominated by atmospheric muons and pure noise triggers, by several orders of magnitude compared to atmospheric neutrino rates. Further levels of event selection are therefore required to obtain a neutrino-dominated event sample. These levels as well as their rates are summarized in Fig. 6.

The IC93 filter is adapted from the DeepCore filter [8] in order to include the new strings. The filter uses a veto region around DeepCore to eliminate muon events which leave light in this region. For each triggered event, the pulses are split into those occurring in the fiducial volume and the veto region. For pulses in the fiducial region, the center of gravity in space and time is calculated. If one of the pulses in the veto region is causally connected to this center of gravity, the event is rejected.

TABLE III. Overview of task-specific subsamples used for training. “Size” here indicates the number of simulated events in the sample. The pulse cleaning subsample uses uncleaned events, whereas all other samples have been cleaned by the GNN-based method.

Task	Size (millions)	Composition
Pulse cleaning	4.04	$T(41\%), C(51\%), \mu(8\%)$
Reco	2.39	$T(46\%), C(54\%)$
T/C	1.81	$T(53\%), C(47\%)$
ν/μ	0.367	$T(27\%), C(32\%), \mu(41\%)$

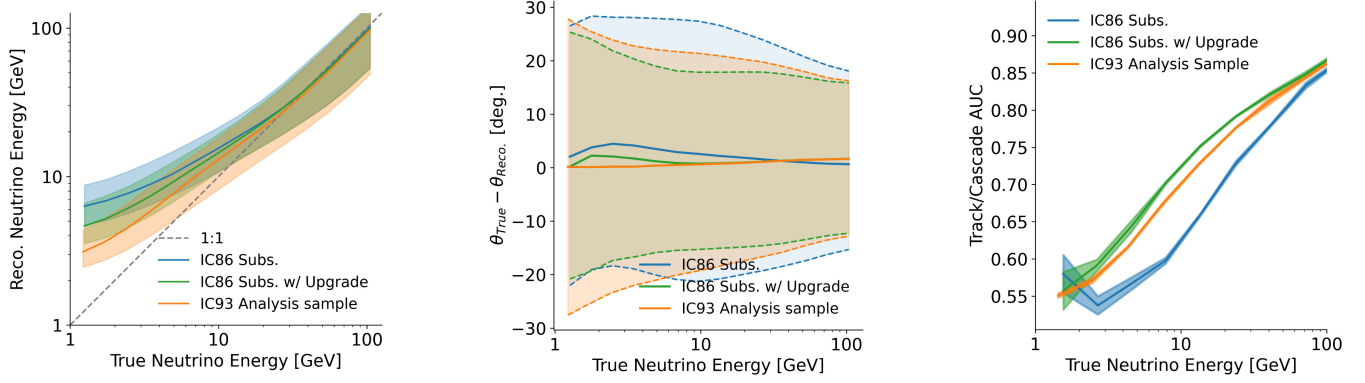


FIG. 5. Performance of reconstruction models on analysis sample (orange) and the IC86 subsample with and without the information on the upgrade strings (green and blue, respectively). Left: reconstructed neutrino energy vs true neutrino energy. Bands depict the 1σ range, and the median is shown in solid. Middle: angular resolution of the zenith reconstruction. Bands represent the 1σ range of the residual distribution around the median shown in solid. Right: area under curve (AUC) score of T/C classifier binned in neutrino energy. Uncertainty bands on the AUC curves result from bootstrapping and represent the expected statistical 1σ deviation. A value of 0.5 indicates random classification, and a value of 1.0 indicates perfect classification. There is a moderate improvement in energy and zenith reconstruction, while the improvement in T/C classification is quite substantial especially around 10 GeV.

Next, a series of simple straight cuts eliminate the most obvious remaining background events. Cuts on five different variables are used. Four of them are based on the cleaned pulse series from the noise cleaning procedure described in Sec. IV B: the number of modules with pulses, the number of pulses in the fiducial volume, the z-position of the first module that saw light, and the time difference between the first and the last pulse. The fifth variable is the time difference between the first and the last pulse before noise cleaning. These variables were selected based on previous experience with DeepCore samples and were also used in [8] with similar cut values. The pure noise events are already suppressed by these cuts by more than three orders of magnitude, mainly by the first two cuts.

Next, a collection of machine learning classifiers is used to identify muon background events. One classifier is the

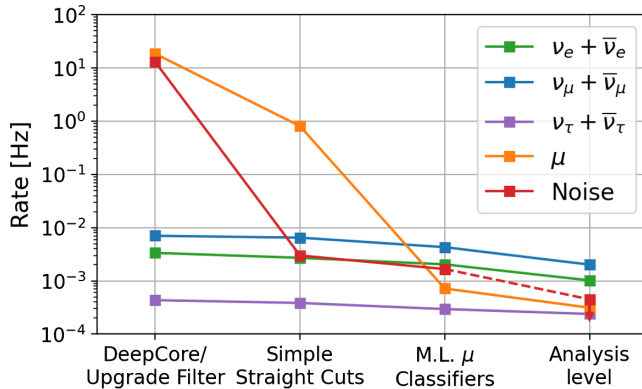


FIG. 6. Signal and background event rates for the different levels of selection. For the noise rate at analysis level, the 90% upper limit is shown (indicated by the dashed line and the arrow). NuFit 5.2 [42] and the normal ordering are used to calculate the neutrino rates.

GNN described in Sec. IV C, the other two are boosted decision trees (BDTs) that use reconstructed quantities and variables describing the spatial and temporal distribution of the pulses. More details about the BDTs can be found in [43]. Applying all three classifiers leads to a reduction of the remaining muon background by three orders of magnitude. Table IV compares the IC93 neutrino rates that we obtain through this selection with the rates from IC86 which use a selection similar to [1] and [8]. The IC93 rates are significantly increased with respect to IC86.

Figure 7 shows the expected event distributions by flavor after the above-mentioned cuts are applied for the three variables used to bin events at the analysis level: reconstructed energy, reconstructed $\cos(\text{zenith})$, and the GNN track score indicating whether an event is more cascadelike or tracklike. Note that the $\nu_e + \bar{\nu}_e$ component in the top panel peaks at reconstructed energies of about 3–5 GeV. These events are crucial in determining the neutrino mass ordering, and the lower the energy threshold of the IceCube Upgrade is key in achieving the increase in event rate in this region. This is a major contributing factor to the improved NMO sensitivity described in more detail in Sec. VI C.

Finally, a few “analysis level” cuts on reconstructed energy and zenith are made to focus on the parameter space

TABLE IV. Neutrino and atmospheric muon rates after event selection (but before final analysis level cuts) for the current (IC86) and upgraded (IC93) detector setting.

Flavor	IC86 (mHz)	IC93 (mHz)
$\nu_e + \bar{\nu}_e$	0.25	2.0
$\nu_\mu + \bar{\nu}_\mu$	0.76	4.3
$\nu_\tau + \bar{\nu}_\tau$	0.05	0.3
μ_{atm}	0.02	0.85

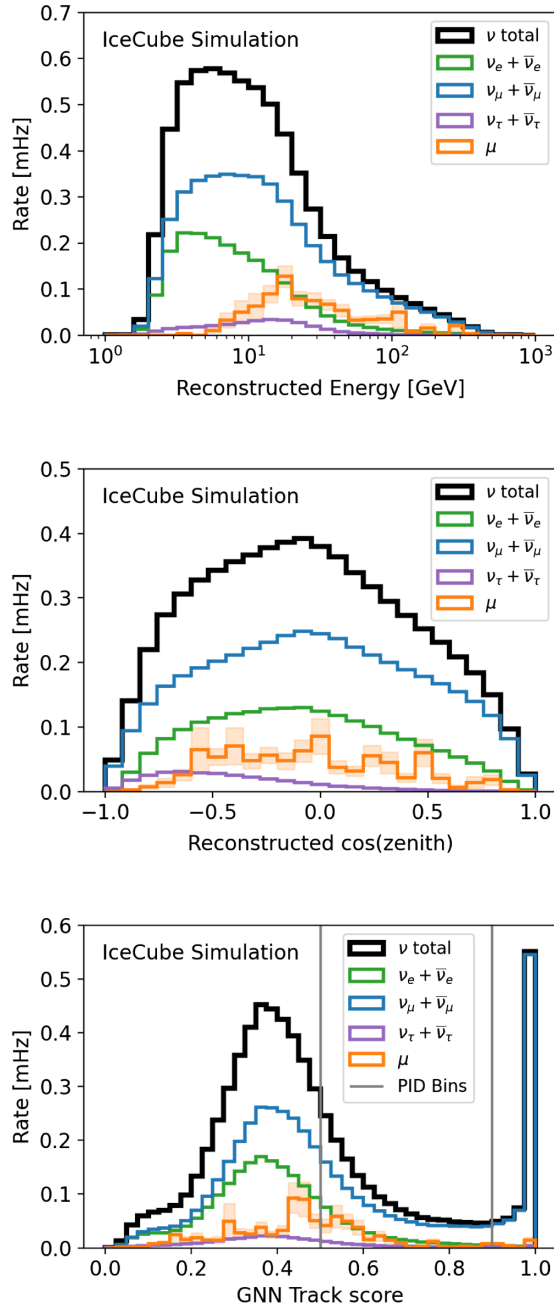


FIG. 7. Expected event distributions by flavor for the three variables used to bin events in the analysis: energy (top), $\cos(\theta_{\text{zen}})$ (middle), and particle identification score (bottom). In the bottom panel, a value of 1 indicates more tracklike events and the gray lines correspond to the boundaries used for particle classification.

where most of the oscillation signal comes from and reduce some of the remaining muon background. This step requires a reconstructed energy between 3 and 300 GeV, and reconstructed $\cos(\text{zenith})$ less than 0. At this level, the event sample is dominated by neutrinos, as can be seen in Fig. 6. None of our simulated pure noise events survive after the final level cuts. Therefore, we show the 90% upper limit of the pure noise rate based on the simulated livetime.

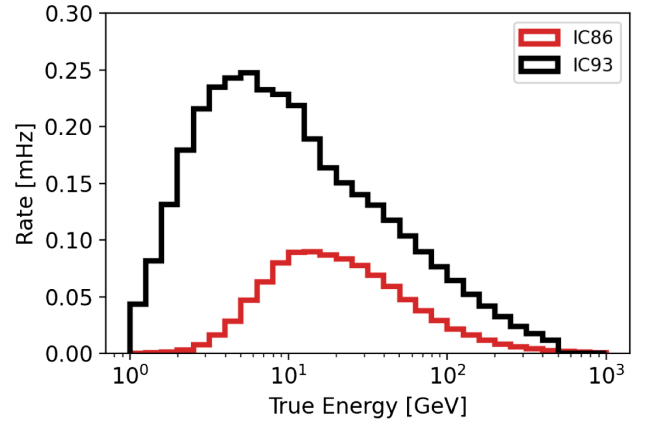


FIG. 8. Comparison of the true energy distribution for simulated neutrino events in the IC86 and IC93 samples. The additional strings significantly increase the number of neutrinos detected.

E. Final level sample

Figure 8 shows the final level energy distributions after all cuts are made, and compares to the energy distribution for the IC86 sample. The energy distribution for IC93 peaks below 10 GeV, whereas for IC86 it peaks above 10 GeV. Additionally, the number of expected neutrino events increases across the entire energy range when using this selection.

V. ANALYSIS

A. Analysis setup

Once the additional strings are deployed there will already be approximately 12 years of data livetime with the current IC86 detector configuration. Therefore we use a combined analysis strategy to incorporate the existing years of IC86 data together with the anticipated IC93 data. Figure 9 shows the expected event distribution with 12 years of IC86 and with 3 years of IC93. The events are binned in energy, $\cos(\text{zenith})$, and particle classification.

The two simulated datasets share almost all simulation tools (Sec. III A) and systematic parameters (Sec. V B) except for the efficiency of the new modules (DOM eff. ICU (IceCube Upgrade), but differ in the noise cleaning, reconstruction methods, event selection, and analysis binning. For IC86 we use the low-level processing described in [8], the likelihood-based event reconstruction presented in [44], and the analysis settings from [31]. The new tools discussed in (Secs. III B–IV D) are applied only to IC93. The treatment of IC86 vs IC93 datasets will be unified in the future.

For the sensitivity projections we compare two scenarios: one with no new hardware installed and data-taking continuing as it is today (denoted as “IC86”). The other scenario includes the combined fit between 12 years of

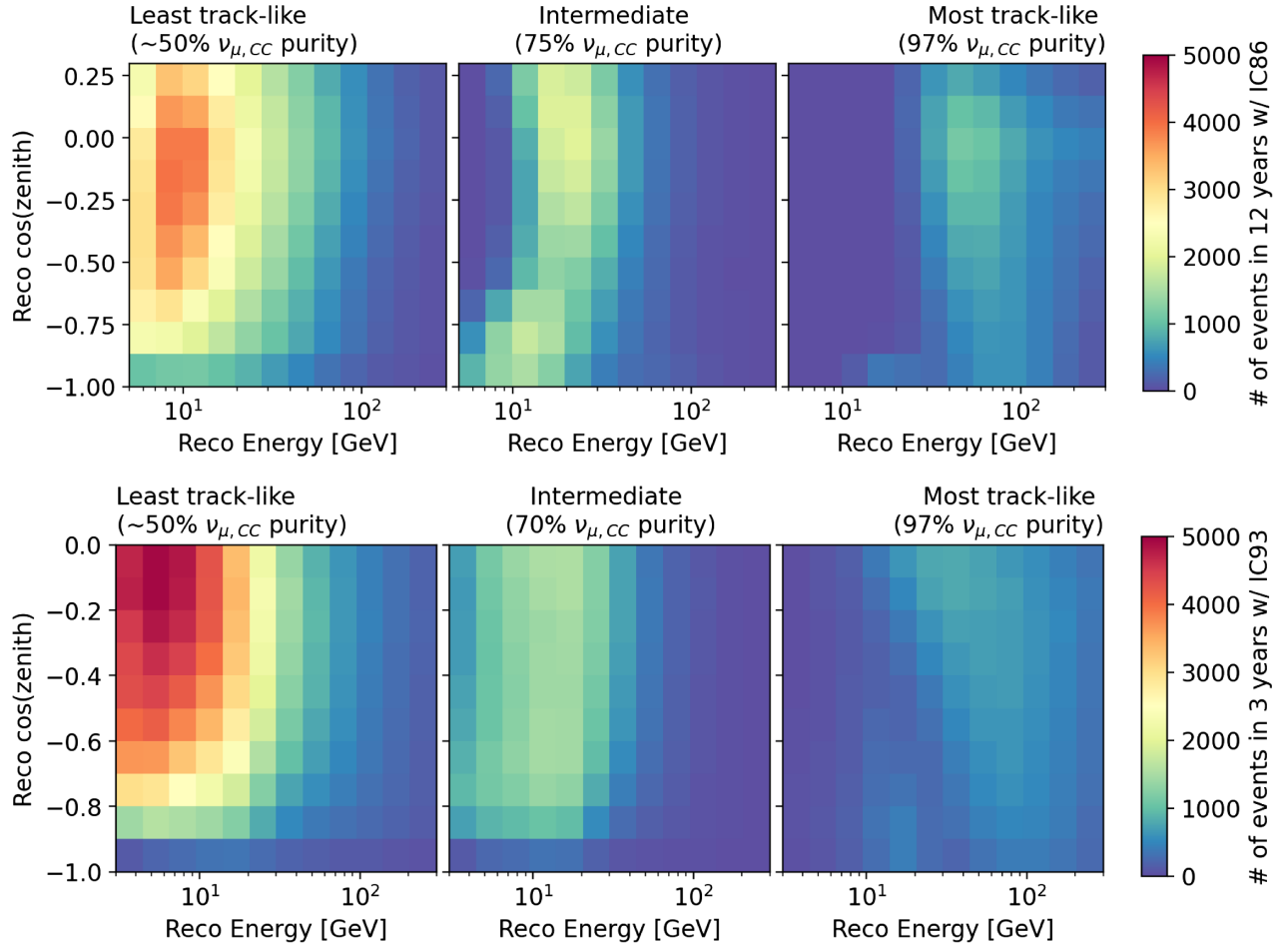


FIG. 9. Distribution of events in the analysis binning used in this study. The top panel shows the events expected with 12 years of the current configuration (IC86), and the bottom panel shows the events expected with 3 years with the additional strings (IC93). Events are binned in energy, $\cos(\text{zenith})$, and particle classification. The binning used in the top panel is the same configuration as existing IceCube DeepCore analyses, while the binning used for IC93 has been reoptimized for the new sample. The number of low energy events is increased with IC93 compared to IC86, even though it assumes fewer years of exposure. The sensitivities presented in this paper combine both of these datasets.

IC86 with additional years using the extra seven strings of the IceCube Upgrade from 2026 onward (denoted as “IC86 (12 yr) + IC93”).

To cover different possible scenarios, we show sensitivities at two different injection points. NuFit 5.2 with and without Super-Kamiokande (SK) data. We perform an Asimov analysis [45] meaning that no pseudotrials are generated to calculate sensitivities, but rather the expected neutrino counts in each analysis bin are used. This means that all sensitivities presented in this paper represent the median sensitivity. A modified χ^2 , which takes into account uncertainties from MC statistics, is used as the test statistic

$$\chi_{\text{mod}}^2 = \sum_i^{\text{bins}} \frac{(n_i - \mu_i)^2}{\mu_i + \sigma_i^2} + \sum_j^{\text{syst}} \frac{(\hat{s}_j - s_j)^2}{\sigma_j^2}. \quad (1)$$

Here n_i is the measured number of events in bin i , μ_i is the expected number of events, and σ_i is the MC

uncertainty. The second term represents the prior penalties for the systematic parameters j which have Gaussian priors σ_j .

B. Systematics treatment

Atmospheric neutrino oscillation experiments have several major categories of systematic uncertainties that are incorporated as nuisance parameters in analyses. They arise from uncertainties in the atmospheric neutrino flux, the neutrino interaction cross sections, and the detector itself. All systematics that were found to be important for the analyses presented in this paper are summarized in Table V. These systematic uncertainties are informed by previous IceCube/DeepCore oscillation analyses.

Flux: The baseline atmospheric neutrino flux spectrum is the model by Honda *et al.* [28]. Atmospheric neutrinos are created in hadronic decays after cosmic-ray interactions in the atmosphere. The uncertainty on the primary cosmic-ray

TABLE V. The parameters included in each of the analyses as physics and/or systematic nuisance parameters, along with the priors assumed for each. Parameters with Gaussian priors list the $\pm 1\sigma$ range and parameters with uniform priors list the full allowed range. The last two columns indicate if the parameter is fit as a nuisance parameter in a given analysis (X for fixed, ✓ for free). A ✓(poi) indicates that the free parameter is the physics parameter of interest. The atmospheric oscillation parameters ($\theta_{23}, \Delta m_{32}^2$) and tau neutrino normalization (ν_τ) analyses use the same list of free/fixed nuisance parameters, whereas the neutrino mass ordering (NMO) analysis has its own set.

Parameter	Nominal	Prior	$\theta_{23}, \Delta m_{32}^2$	ν_τ	NMO
<i>Detector:</i>					
DOM efficiency IC86	1.0	± 0.1	✓	✓	✓
DOM efficiency ICU	1.0	± 0.05	✓	✓	✓
Ice absorption	1.0	± 0.05	✓	✓	✓
Ice scattering	1.0	± 0.1	✓	✓	✓
Relative efficiency p_0	0.10	$[-0.6, 0.5]$	✓	✓	✓
Relative efficiency p_1	-0.05	$[-0.15, 0.05]$	✓	✓	✓
<i>Flux:</i>					
$\Delta\gamma_\nu$	0.0	± 0.1	✓	✓	✓
$\Delta\pi^\pm$ yields D	0.0	± 0.3	X	X	✓
$\Delta\pi^\pm$ yields G	0.0	± 0.3	✓	✓	✓
$\Delta\pi^\pm$ yields H	0.0	± 0.15	✓	✓	✓
$\Delta\pi^\pm$ yields I	0.0	± 0.61	✓	✓	✓
ΔK^+ yields W	0.0	± 0.4	✓	✓	✓
ΔK^+ yields Y	0.0	± 0.3	✓	✓	X
ΔK^+ yields Z	0.0	± 0.122	X	X	✓
<i>Cross section:</i>					
M_A^{CCQE} (in σ)	0.0	± 1.0	✓	✓	✓
M_A^{CCRES} (in σ)	0.0	± 1.0	✓	✓	✓
M_A^{NCRES} (in σ)	0.0	± 1.0	✓	✓	✓
M_A^{coh} (in σ)	0.0	± 1.0	X	X	✓
DIS CSMS	0.0	± 1.0	✓	✓	X
ν_τ x-sec scale	0.0	$[-1.0, +1.0]$	X	X	✓
<i>Normalization:</i>					
A_{eff} scale	1.0	$[0.1, 2.0]$	✓	✓	✓
<i>Atmospheric muons:</i>					
Atmospheric μ scale	1.0	$[0.1, 3.]$	✓	✓	✓
<i>Oscillations:</i>					
θ_{13}	NuFit 5.2	0.11	X	X	✓
θ_{23}	NuFit 5.2	None	✓ (poi)	✓	✓
Δm_{32}^2	NuFit 5.2	None	✓ (poi)	✓	✓
ν_τ normalization	1.0	None	X	✓ (poi)	X
Mass ordering	Normal	Not applicable	X	X	✓ (poi)

spectrum is modeled with a power-law correction $E^{\Delta\gamma_\nu}$ [46]. To incorporate uncertainties on the hadronic interactions, the MCEq package [47], which computes fluxes of atmospheric leptons, was used to calculate correction gradients that resemble the sources of systematic uncertainty from [46]. Depending on the primary energy and the fraction of energy taken by the secondary meson, different uncertainties are used. While there are 17 parameters for the variations of the yields of $\pi^\pm$ and $K^\pm$, only 7 were found to be important for the analyses presented in this paper. They are listed in Table V.

Cross section: We simulate neutrinos with energies ranging from 1 GeV up to 500 GeV. Above ~ 20 GeV, deep inelastic scattering (DIS) is the dominant interaction type. We use GENIE [27] and Cooper-Sarkar-Mertsch-Sarkar (CSMS) [48] to calculate the DIS cross section and interpolate between the results. We use the parametrization described in [8], where DIS CSMS = 0 means that the GENIE value is used, and DIS CSMS = 1 would approximate the CSMS prediction. At lower energies, resonance production (RES), quasielastic scattering (QE), and coherent scattering (coh) become more

important. In GENIE it is possible to reweight the cross sections of these interactions based on a change in the axial mass (M_A).

For the neutrino mass ordering analysis (see Sec. VIC), we also include a parameter that scales between different ν_τ cross section models. A value of 0 indicates that the GENIE value is used; a nonzero value indicates a correction has been applied to reweight events using a band that covers a range of models from [49]. This parameter is not included in the ν_τ -norm study (see Sec. VIB) because the ν_τ normalization can itself be interpreted as cross section measurement; instead, the parameter is fixed to GENIE.

We have also studied the effect of differential DIS cross-section corrections to GENIE to better describe NuTeV data [50]. These corrections are applied to the event weights, modeled as ax^{-b} where x is the Bjorken- x , and a and b are fit to the data. This was done independently for neutrinos and antineutrinos. The effective parameters were extracted using the NuTeV dataset at $E_\nu = 35$ GeV. Since a fit to the $E_\nu = 65$ GeV NuTeV dataset yielded compatible parameter values, we assume no energy dependence. The corrections were found to have only a minor impact on the physics parameters and were mostly absorbed by our existing GENIE and flux systematics, and were therefore not explicitly included in the study.

In addition, we tested a similar procedure to model the impact of changes to the DIS Bodek-Yang correction [51]. Here the functional form $ax^{-b}y^{-c}$ was used where x is the Bjorken- x and y is the Bjorken- y or inelasticity. This parametrization showed similarly small impact on the physics parameters.

Detector: Uncertainties in the modeling of the optical properties of the ice and the efficiency of the optical modules have the greatest impact on our analyses. Our baseline ice model is described in [35]. Two scaling factors are used to vary the strength of the photon absorption and scattering in the ice, respectively. The optical properties of the refrozen ice in the immediate vicinity of the strings, which are altered by air bubbles and other impurities, are modeled following [36]. This leads to two additional parameters (p_0 , p_1) describing the modified angular efficiency of the optical modules. Note that this is only modeled for existing DOMs. During the deployment of the upgrade, a modified drilling process is used to reduce air bubbles and other impurities. This is expected to improve the clarity of the refrozen ice surrounding the new modules. The uncertainty on the overall optical efficiency of the existing models (DOM eff. IC86) as well as the new modules (DOM eff. ICU) is taken into account. All new modules share the same efficiency scale, which has a tighter constraint than the one for the existing DOMs. Thanks to the new calibration devices, we assume the new modules to be calibrated with higher precision; the new modules have an uncertainty of 5% and the existing modules have an

uncertainty of 10% as indicated in Table V. To implement the effect of the detector systematics, for IC86 we use the “hypersurface” method described in [8], while for IC93 we use the likelihood-free inference method from [52]. In future studies, both detector configurations will use the same method.

Normalization: An overall normalization is applied to the neutrinos (A_{eff} scale) and the atmospheric muon background (Atm. μ scale) separately.

Earth model: For the matter density profile of the Earth, the preliminary reference Earth model [53] is used. For simplification, we assume 12 radial layers of fixed density, which does not lead to any additional free nuisance parameters.

There are more systematics that are being targeted as part of the upgrade ice calibration, such as the exact geometry/position of the modules after deployment, the optical properties of the refrozen ice directly surrounding the modules, uncertainties on the depth and direction-dependent photon scattering function. They are not considered in this paper, but will be studied after the deployment using the calibration devices mentioned in Sec. IIC.

VI. EXPECTED SENSITIVITIES

A. Sensitivity to θ_{23} and Δm_{32}^2

The mixing between neutrino states is very strong, with θ_{23} being close to maximal mixing. This makes it important to measure the neutrino oscillation parameters, especially θ_{23} , because they can reveal structure in the weak flavor mixing of leptons. To obtain the expected constraints on the atmospheric oscillation parameters Δm_{32}^2 and θ_{23} , we follow a similar analysis procedure as for the existing IceCube measurements [1,8].

Figure 10 shows the sensitivity at the 90% confidence level with 3 years of IC93 plus 12 years of IC86 data. To illustrate the sensitivity gain from the additional strings, we also show a scenario with 15 years of data from the current detector configuration (IC86, dotted lines). The left plot assumes NuFit 5.2 [42] (w/o SK data) as injected truth, while the right plot uses NuFit 5.2 (w/ SK).

Thanks to the additional strings, IceCube’s 90% CL region for Δm_{32}^2 and θ_{23} shrinks by about 70% (w/o SK) and 55% (w/ SK). IceCube’s ability to rule out maximal mixing or determine the octant strongly depends on the true value of θ_{23} and how far away it is from maximal. As seen in Fig. 10, for an injected truth given by the NuFit 5.2 w/o SK value in which θ_{23} is farther from maximal, we could exclude maximal mixing at the 90% confidence level, and the addition of the IceCube Upgrade strings could allow us to rule out the lower octant. However, for other injected truth values like NuFit 5.2 w/ SK value, the octant is not resolved and maximal mixing is not able to be rejected at the 90% confidence level. However this study was not specifically optimized for resolving the octant or rejecting

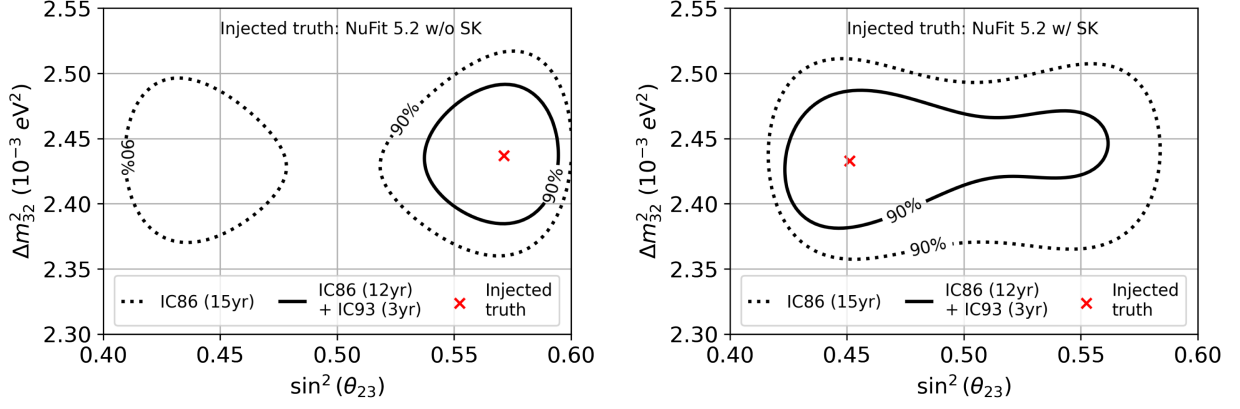


FIG. 10. Expected 90% confidence level sensitivity contours for the atmospheric oscillation parameters 3 years after the planned deployment of the additional strings. A scenario w/ (solid) and w/o (dotted) the additional strings is compared in each plot. The left plot uses NuFit 5.2 w/o SK as injected truth, while the right plot assumes NuFit 5.2 w/ SK.

maximal mixing so future work could improve IceCube's sensitivity to these questions.

Appendix A shows one-dimensional projections of these sensitivities, as well as the two-dimensional contour for an additional scenario where the injected truth is given by [1]. Our projected 3 year sensitivities compared to current experimental measurements [1,2,54–57] are shown in Fig. 11. The full χ^2 maps used to produce these sensitivities are publicly available in [58].

B. Sensitivity to ν_τ normalization

We will also measure the overall rate of ν_τ and compare it to the oscillation expectation. This is done by scaling the number of observed ν_τ by a normalization factor and can be interpreted as test of the unitarity of the PMNS matrix as well as a test of the ν_τ cross section. A normalization value of 1.0 indicates the nominal expectation. A value other than 1.0 could hint at new physics. More information about how IceCube measures ν_τ appearance can be found in [59].

Figure 12 illustrates our sensitivity to the ν_τ normalization for a true value of 1.0. The left plot shows the rejection power for different ν_τ normalization values assuming data from 3 years after the planned deployment, while in the right plot the 1σ uncertainty on ν_τ normalization is shown for different detector operation times. Since the choice of injected oscillation parameters has a negligible impact on this analysis, we only show the results for NuFit 5.2 w/o SK.

Already after a few years the uncertainty on the ν_τ normalization can be reduced by almost a factor of two compared to the configuration without the additional strings. A precision of 5% at 1σ is expected with 3 years of livetime. Future studies could constrain nonunitarity more directly by fitting the PMNS matrix elements, similar to the method in [60].

C. Sensitivity to neutrino mass ordering

This analysis covers our sensitivity to the ordering of the three neutrino mass eigenstates, which can either be

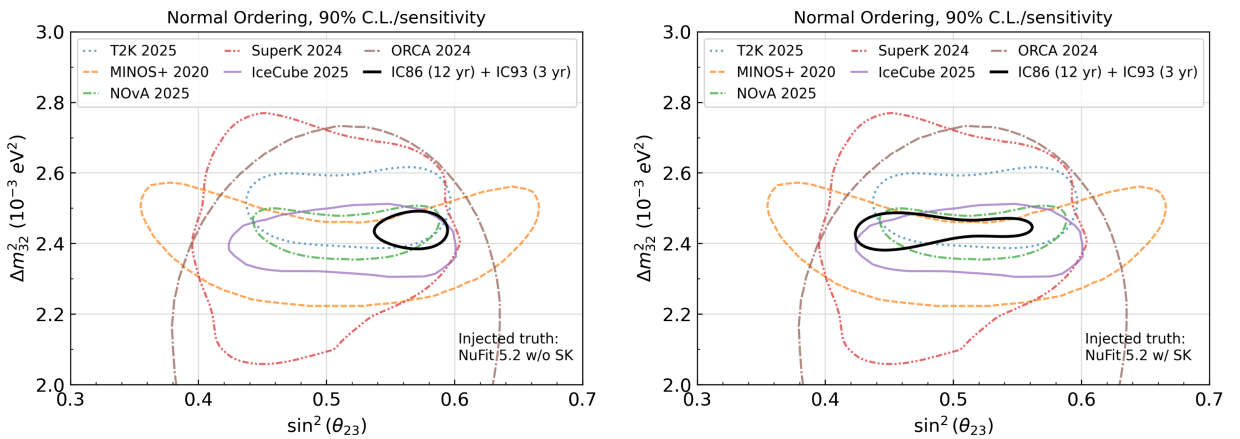


FIG. 11. The expected IceCube 90% confidence level sensitivity contour with 3 years of livetime including the additional strings compared to existing measurements of the atmospheric oscillation parameters. The left plot uses NuFit 5.2 w/o SK as injected truth, while the right plot assumes NuFit 5.2 w/ SK. Results from several experiments are shown for comparison [1,2,54–57].

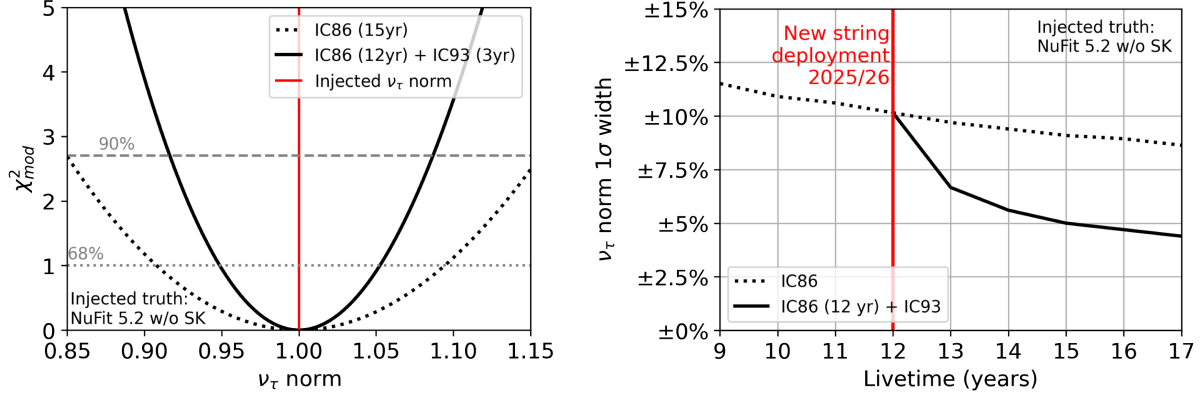


FIG. 12. Expected sensitivity for constraining the ν_τ normalization scaling factor. The left plot shows a profile scan 3 years after the planned deployment of the additional strings, while the right plot shows the 1σ width of such a scan over detector livetime. In both plots a scenario w/ (solid) and w/o (dotted) the additional strings is compared.

“normal” or “inverted.” Recent cosmological measurements of baryonic acoustic oscillations provide an upper limit on the sum of neutrino masses that is remarkably close to the minimum allowed value from neutrino oscillations for the inverted ordering. Therefore, a direct measurement of the NMO is of great importance: the inverted ordering could be an indication of the need to adjust cosmological models or the reconsideration of standard neutrino oscillation assumptions [61].

All sensitivities refer to median sensitivities as defined in [5]. More information about how IceCube determines the NMO can be found in [62]. As shown in Table V, the selection of free parameters was adjusted for this analysis, because the signal region for this analysis is at lower energies compared to the analyzes presented in Secs. VIA and VIB.

Figure 13 shows the expected median NMO sensitivity evolution as a function of the detector livetime assuming a true normal ordering (left plot) or a true inverted ordering

(right plot). Since the NMO sensitivity of atmospheric neutrino experiments strongly depends on the true value of θ_{23} , Fig. 13 does not provide a single median sensitivity but a range (shaded areas) of possible sensitivities depending on the assumed true value of θ_{23} . In addition, we show sensitivities for three example values of θ_{23} : the value from NuFit 5.2 w/o SK (solid line) in which θ_{23} is in the upper octant, NuFit 5.2 w/ SK (dashed line) in which θ_{23} is in the lower octant, and the IceCube result from [1] (dotted line). In contrast to accelerator long-baseline experiments, a CP violating phase δ_{CP} in the mixing matrix has negligible impact on our NMO sensitivities. This is mainly due to the higher energy range of atmospheric neutrinos.

The sensitivity improvement from the additional strings can be seen by comparing the gray and red sensitivities right of the vertical gray line, which marks the deployment of the additional strings. The improvement in sensitivity results from both the increase in the number of low energy events and from the improved reconstruction resolutions.

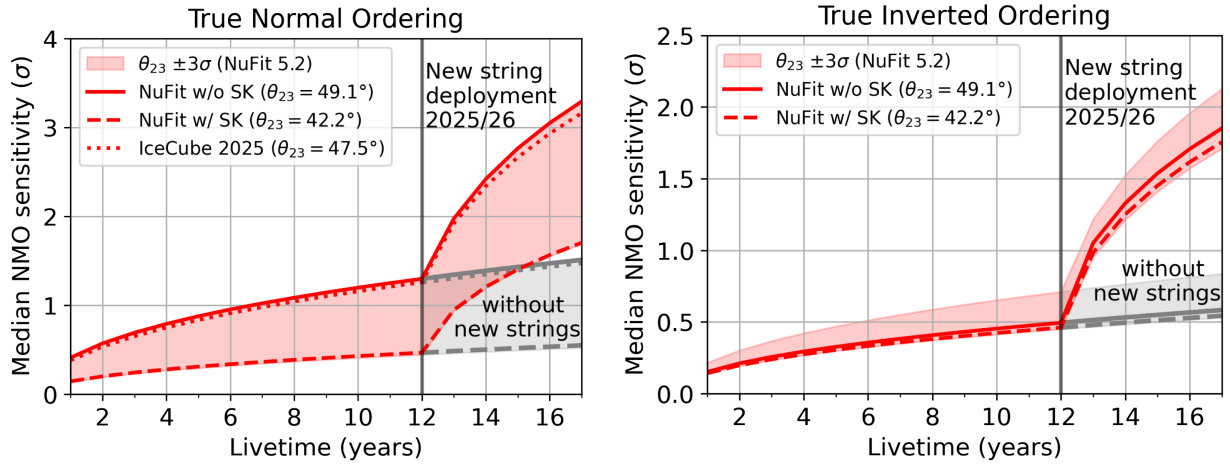


FIG. 13. Livetime evolution of the median NMO sensitivity for different true values of θ_{23} . The shaded region marks the possible range of sensitivities for the NuFit 5.2 3σ range of θ_{23} . The left plot is for a true normal ordering, and the right plot for a true inverted ordering.

For a true normal ordering a detection significance of up to 3σ is possible within 5 years, while for a true inverted ordering up to 2σ is possible.

VII. CONCLUSION

Seven additional strings, hosting newly developed optical modules, will be deployed for the IceCube Upgrade in the Polar season of 2025–2026. These strings will significantly increase the number of GeV neutrino interactions observed by the detector, as well as the ability to reconstruct and classify events. These new strings will therefore lead to significant improvement in our sensitivities to many neutrino oscillation analyses, particularly the three analyses considered in this paper. With three years of data from the upgraded detector, we expect to collect about 3.3×10^5 neutrino events with the event selection presented here. Combining the new data with the approximately 2.8×10^5 neutrinos expected from the 12 year data taken by the current IceCube DeepCore, yields the sensitivities presented in this paper. Compared to a scenario without the additional strings, we see an improvement of 55%–70% in the area enclosed by the 90% CL contour for the atmospheric neutrino oscillation parameters, around 40% in the 1σ range for the ν_τ normalization, and a factor of $2 - 3\times$ boost in the median NMO sensitivity.

Additional optimizations of triggers, event selection, processing, reconstructions, analysis choices, and models for treating systematic uncertainties are expected to further improve the sensitivities presented here. In addition, opportunities for future improvements in sensitivity can be achieved through combined fits with other experiments. For example, previous studies have shown that a combined fit between the IceCube Upgrade and a medium baseline reactor neutrino experiment like Jiangmen Underground Neutrino Observatory (JUNO) [63] will significantly increase the joint NMO sensitivity [64].

Beyond atmospheric oscillations, many additional analyses will also benefit from the reduced energy threshold, higher event rate, and improved precision of the IceCube Upgrade. Examples for such analyses are dark matter searches, beyond the standard model searches, and GeV neutrino astronomy. Finally, the extensive suite of calibration devices that will be deployed in the upgrade will lead to better knowledge of the ice properties and reduce the impact of detector systematic uncertainty across all IceCube analyses. The IceCube Upgrade is expected to bring unprecedented precision to the atmospheric neutrino sector.

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DATA AVAILABILITY

The data that support the findings of this article are openly available [58].

APPENDIX A: ADDITIONAL SENSITIVITIES

Figure 14 shows the 1D projections of the expected sensitivity to the atmospheric oscillation parameters $\sin^2(\theta_{23})$ and Δm_{32}^2 . These correspond to the 2D projections shown in Figs. 10 and 11.

In addition to computing the sensitivity for scenarios provided by global fits (Figs. 10 and 11), Fig. 15 shows the expected sensitivity for a scenario where the true value

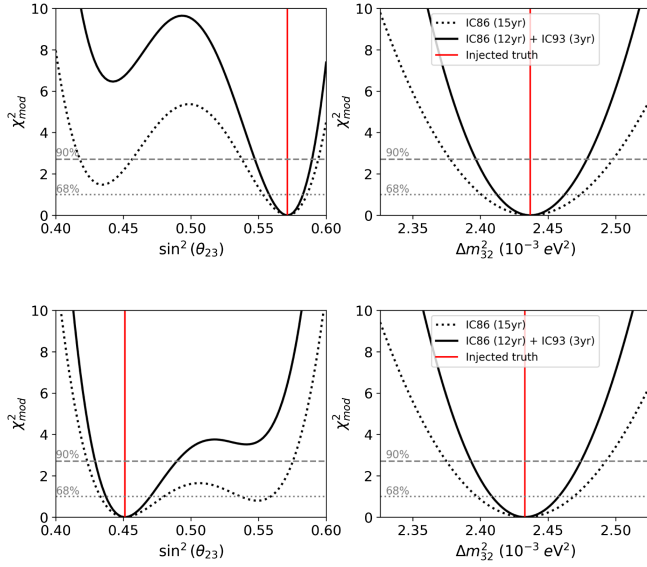


FIG. 14. Expected sensitivity to the atmospheric oscillation parameters with 3 years of livetime with the new strings. A scenario w/ (solid) and w/o (dotted) the additional strings is compared in each plot. The top plots use NuFit 5.2 w/o SK as injected truth, while the bottom plots assume NuFit 5.2 w/ SK.

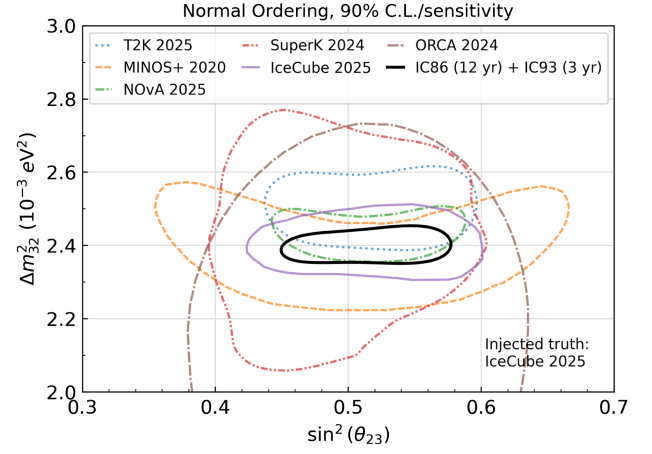


FIG. 15. Expected 90% confidence level contours for the atmospheric oscillation parameters with 3 years of livetime with the new strings, where the assumed truth is given by [1].

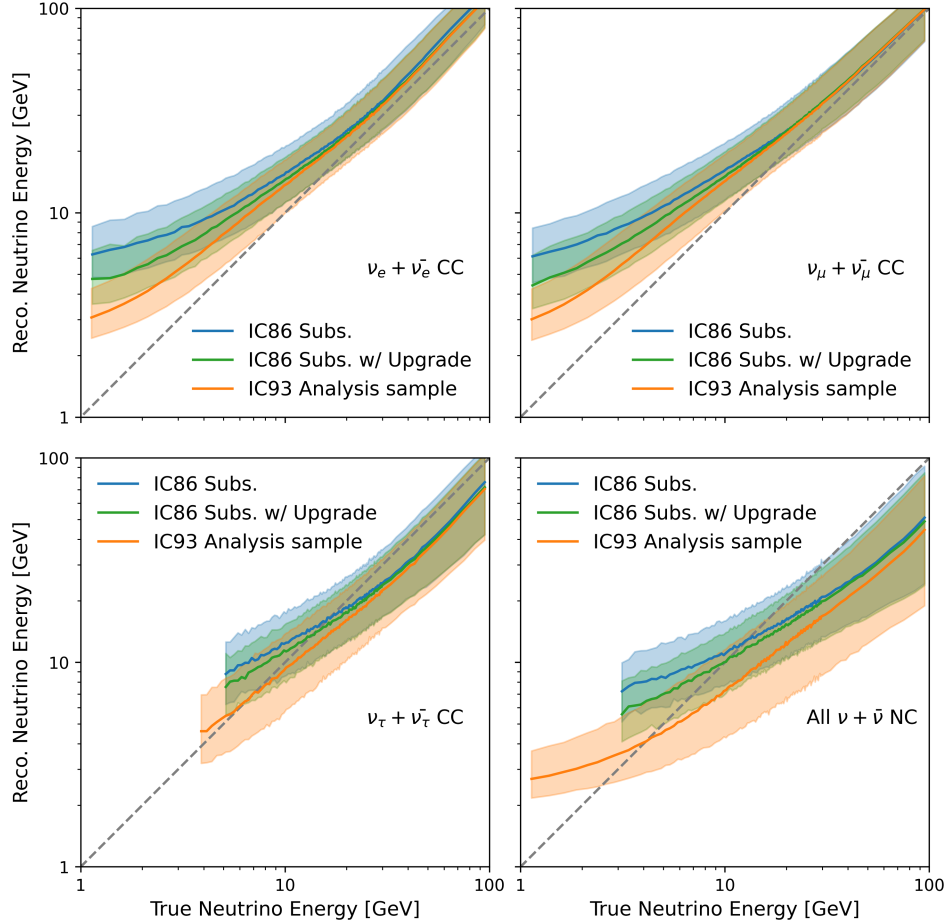


FIG. 16. Energy resolution per flavor type for three scenarios: the full IceCube Upgrade analysis sample (orange) and the IC86 subsample with and without the information from the IceCube Upgrade strings (green and blue respectively).

realized in nature is the best-fit point as measured by the most recent IceCube DeepCore measurement [1].

APPENDIX B: RECONSTRUCTION COMPARISONS

The analysis sample presented in Sec. IV includes low energy events that would not have been detectable by the existing IC86 detector configuration. Therefore comparing reconstruction resolutions for existing IC86 samples with the new simulated IC93 sample does not offer a direct comparison since the underlying distribution of event types has changed. A more direct comparison of the expected improvement in reconstruction resolutions due to the new strings requires a subset of events from the new IC93 simulated sample that would also have been detected by the IC86 detector configuration.

The full sample described in Sec. IV is denoted here as “IC93 Analysis Sample.” We then identify the subsample of events that are detectable by IC86 and could be improved with the inclusion of the new strings, defined as events with at least 8 pulses in the IC86 array and at least 1 pulse in the IceCube Upgrade extension. This subsample is denoted

“IC86 Subsample.” These events are reconstructed with and without the pulses on the new IceCube Upgrade strings. The reconstruction methods are retrained for each separate instance. By comparing the performance of the reconstruction methods on the exact same events, with and without the information from the Upgrade strings, we can directly compare the effect of the new hardware on reconstruction performance. Figures 16 and 17 quantify the difference in reconstruction for the three scenarios: the full IceCube Upgrade analysis sample (“IC93 Analysis Sample”) and the IC86 subsample with and without the information from the Upgrade strings (“IC86 Subs. w/ Upgrade” and “IC86 Subs.” respectively).

Compared to Fig. 5, these figures are further subdivided by neutrino flavor and interaction channel. In the top panels of Fig. 16, it can be observed that $\nu_e + \bar{\nu}_e$ CC and $\nu_\mu + \bar{\nu}_\mu$ CC have both improved variance and bias compared to NC interactions (bottom right of Fig. 16). This difference is expected as the outgoing neutrino in NC interaction escapes the detector with parts of the energy of the incident neutrino, leading to a systematic underestimation of the energy of the incident neutrino. In general, the information about the $\nu_\mu + \bar{\nu}_\mu$ CC energy

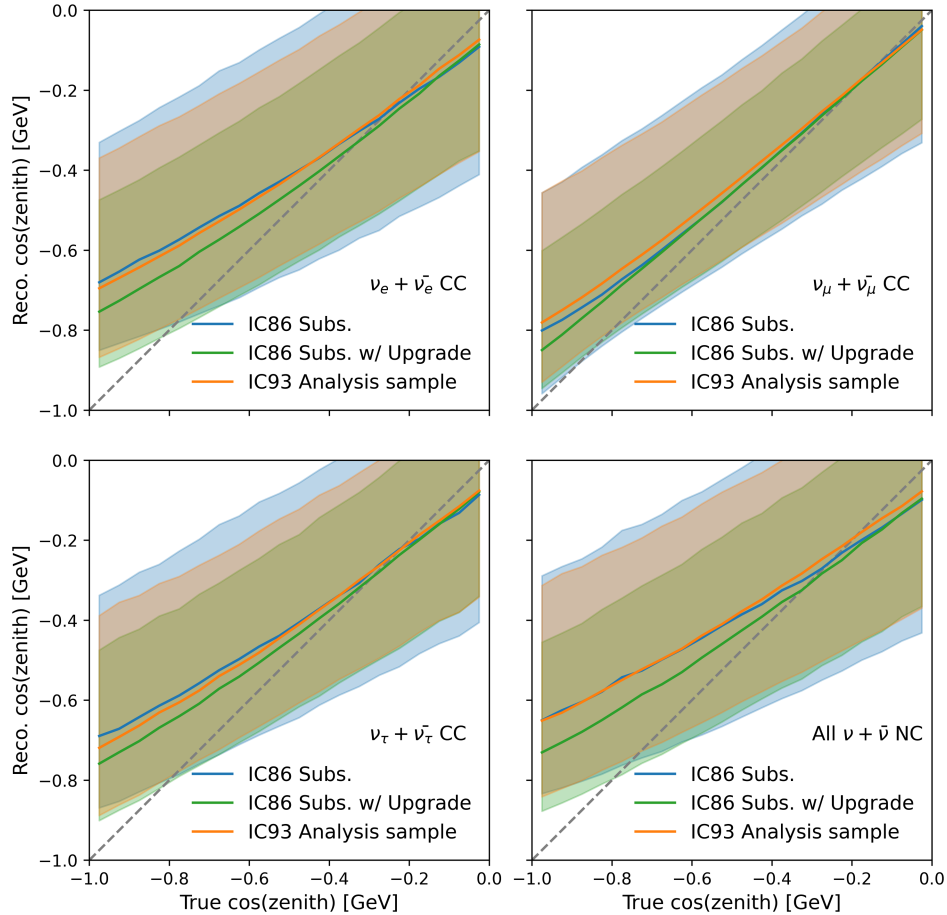


FIG. 17. Zenith resolution per flavor type for three scenarios: the full IceCube Upgrade analysis sample (orange) and the IC86 subsample with and without the information from the IceCube Upgrade strings (green and blue respectively).

largely comes from the length of the muon track. Similarly, Fig. 17 contains the angular resolutions shown in Fig. 5 subdivided into neutrino flavor and interaction channels. By comparing the blue and green curves in Fig. 17, it can be observed that events in the IC86 subsample have significantly improved variance in zenith

reconstructions and improved bias for up-going events for the scenario when information from the IceCube Upgrade strings is included in the reconstruction. For example, for up-going neutrino events below 50 GeV, the variance (i.e. the width of the blue/green band) is lowered by 24% from 41 degrees (blue) to 31 degrees (green).

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