Measuring atmospheric neutrino oscillations with KM3NeT/ORCA6

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The KM3NeT Collaboration is constructing two water-Cherenkov neutrino detectors at the bottom of the Mediterranean Sea: ARCA, which is designed for neutrino astronomy in the TeV to PeV range, and ORCA, optimised for GeV neutrino detection. The ORCA detector will comprise 115 string-like vertical Detection Units arranged in a cylindrical array. Its main objectives are to determine the neutrino mass ordering and measuring atmospheric neutrino oscillations. During 2020 and 2021, an early configuration of the detector with six lines was in operation. A high-purity neutrino sample covering 433 kton-years of exposure was extracted using optimised reconstruction algorithms and machine learning classifiers. In this contribution, the measurement of the neutrino oscillation parameters $\sin^2 \theta_{23} = 0.51^{+0.06}_{-0.07}$ and $\Delta m^2_{31} = 2.14^{+0.36}_{-0.25} \cdot 10^{-3}$eV$^2$, as well as the sensitivity to determine the neutrino mass ordering based on this data sample, will be presented.
1. Introduction

KM3NeT is a research infrastructure for neutrino experiments located on the seabed of the Mediterranean Sea and currently undergoing construction [1]. The infrastructure comprises two detectors that use the same technology but are designed to achieve distinct physics objectives based on their respective spatial configurations. ARCA (Astroparticle Research with Cosmics in the Abyss) is being installed at the KM3NeT-It site, situated 100 km off the coast of Sicily near Capo Passero, Italy, at a depth of 3500 m. ARCA is dedicated to the search for high-energy neutrinos originating from astrophysical sources. ORCA (Oscillation Research with Cosmics in the Abyss) is being built near the coast of Toulon, France, positioned 40 km offshore at a depth of 2500 m. The primary objective of ORCA is to investigate the Neutrino Mass Ordering (NMO) by detecting the neutrino flux generated in the Earth’s atmosphere [2]. With a planned instrumented volume of about one cubic kilometer of seawater, ARCA will encompass a mass of one gigaton, while ORCA will cover approximately 7 megatons.

The detection system of ORCA consists of Digital Optical Modules (DOMs) which are pressure-resistant glass spheres housing 31 photomultiplier tubes (PMTs) and corresponding readout electronics [3]. These DOMs are arranged along vertical flexible strings called Detection Units (DUs), anchored to the sea floor and maintained in a vertical position through the buoyancy of the DOMs and a submerged buoy at the top. ORCA will comprise 115 DUs, each equipped with 18 DOMs, with a vertical spacing of 9 m and a horizontal separation of about 20 m. Currently, 18 DUs have been deployed, while the presented results correspond to the initial data acquired using the configuration with 6 DUs referred to as ORCA6.

The ORCA detector employs the Cherenkov effect as its detection principle. Charged particles exceeding the local speed of light induce Cherenkov radiation, which is then registered by the DOMs. This mechanism allows for the reconstruction of interaction parameters, including the interaction vertex, and event topology. By harnessing these capabilities, ORCA enables the exploration of the neutrino properties. Notably, the resulting event topology exhibits distinct light emission characteristics appearing as track-like patterns for GeV muons produced in $\nu_\mu$-CC interactions and shower-like patterns for other neutrino channels.

2. Data taking and selection

The ORCA detector has been continuously measuring in the deep sea since mid-2019, while it was extended. From January of 2020 to November 2021, the detector was taking data with a 6-DU configuration. A run selection has been applied to ensure strict data quality criteria resulting in a total of 510 days of data. The time distribution of the exposure is shown in figure 1 in terms of kton-years, taking into account the instrumented volume of seawater, which is corrected by removing non-working PMTs and PMTs with very high rates induced by environmental optical background. The current dataset corresponds to 433 kton-years. A previous preliminary analysis was carried out with a part of this dataset, in particular using a sample equivalent to 296 kton-years [4]. Several improvements have been made in the selection and analysis with respect to the previous analysis: shower reconstruction was added, selection is now based in machine learning algorithms and energy reconstruction was improved.
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The overwhelming majority of recorded events in ORCA are atmospheric muons and pure noise, which have to be rejected in the selection process. To remove these events, first selection is applied by requesting a direction reconstruction with a high quality score, a number of triggered hits above a certain threshold and the reconstructed direction to be up-going, since no atmospheric muons are expected in the up-going direction. A Boosted Decision Tree (BDT) machine learning algorithm is then used to assign particle identification scores to the events that allow to remove remaining atmospheric muons and to discriminate between the two possible event topologies: track-like events and shower-like events. To this end, two scores are calculated which will be called atmospheric muon score and track score, respectively.

After implementing that pre-selection and excluding eventswith high muon scores, the remaining events are initially categorised into two classes based on the track score: tracks and showers. Subsequently, the track class is further divided into two subclasses using the muon score, which serves as a quality parameter to distinguish well-reconstructed tracks from poorly reconstructed tracks that may be misidentified as atmospheric muons. As a result, three distinct classes of events are formed: High Purity Tracks, Low Purity Tracks, and Showers.

To refine the event selection, an additional selection is applied to the reconstructed energy, but with different thresholds for the shower and track classes. Events with a reconstructed energy exceeding 1 TeV for showers and above 100 GeV for tracks are removed. This specific threshold is employed to mitigate the impact of the migration of high-energy events (above 10 TeV) that are not simulated. Based on this selection process, a total of 5830 events are expected from Monte-Carlo (MC) simulations to meet the criteria while a total of 5828 events were actually observed.

Figure 2 displays the compatibility of the data and MC simulations for the described selection process for both scores. Pre-selection is applied to both data and MC distributions. The additional cut on the muon score to reject the atmospheric muon background is already applied in the track score distribution.

The MC distributions are constructed using events generated with gSeaGen [5] and MUPAGE [6], followed by propagation, triggering, and reconstruction using custom KM3NeT software. MC distributions are simulated with oscillation parameters based on NuFit 5.0 [7] (with Super-Kamiokande data) and normal ordering (NO). The measured data and the modelled MC exhibit
good agreement for all the distributions.

3. Analysis

The analysis is based on 2-dimensional distributions of the reconstructed energy and reconstructed cosine of the zenith angle for each of the three event classes. These distributions are obtained through the implementation of Swim [8], an analysis framework developed for KM3NeT which employs MC simulations to model the detector response. The true energy and zenith angle distributions are calculated for each (anti)neutrino interaction type by taking into account cross sections, neutrino fluxes, the interaction volume, and the oscillation probabilities.

Neutrinos are simulated across a wide true energy range, from 1 GeV to 10 TeV, and from all directions. However, only reconstructed up-going directions are used for the analysis, with 10 bins in the reconstructed cosine of the zenith angle. The range of reconstructed energies spans from 2 to 1000 GeV, employing 15 non-equally spaced bins chosen in a way to ensure sufficient statistics per bin. Note that the 15th bin, ranging from 100 to 1000 GeV, exclusively contains shower events, as tracks with reconstructed energies above 100 GeV are rejected.

To account for the detector resolution, a response matrix which is evaluated by reconstructing MC events is employed. This matrix establishes the relationship between the true and reconstructed variables utilised in the analysis. For each interaction channel $\nu_x$ and classification $i$, a 4-dimensional response matrix $R^{[\nu_x \rightarrow i]}(E_{\text{true}}, \theta_{\text{true}}, E_{\text{reco}}, \theta_{\text{reco}})$ is defined. Each entry within the matrix represents the detection efficiency, classification, and reconstruction probability for a given bin of true energy and zenith angle $(E_{\text{true}}, \theta_{\text{true}})$. The number of reconstructed events for a given class $i$ is determined by multiplying the expected number of interacting events with these corresponding efficiencies.

$$n_{\text{reco}}^i(E_{\text{reco}}, \theta_{\text{reco}}) = \sum_x n_{\text{int}}^x(E_{\text{true}}, \theta_{\text{true}}) \times R^{[\nu_x \rightarrow i]}(E_{\text{true}}, \theta_{\text{true}}, E_{\text{reco}}, \theta_{\text{reco}}).$$

The resulting effective mass for different neutrino flavours and interaction types is shown in figure 3 as a function of neutrino energy. Selection cuts are applied. The effective masses continue
Figure 3: Effective mass for different neutrino flavours and interaction types as a function of the true neutrino energy. Selection cuts are applied. The ORCA6 instrumented mass is shown with a vertical line.

growing as energy grows due to the contribution of events starting outside of the instrumented volume.

The analysis procedure to constrain the oscillation parameters is based on the maximisation of a binned likelihood for the 2-dimensional distribution of events in $\log_{10}(E_{\text{reco}}/\text{GeV})$ and $\cos \theta_{\text{reco}}$, comparing the observed data to a model prediction. The sensitivities are computed using the Asimov approach, in which the observed data is replaced by a representative dataset using the expected values of the null hypothesis for each bin [9]. This analysis is not sensitive to $\theta_{13}$, $\theta_{12}$, $\Delta m^2_{31}$ and $\delta_{CP}$, so they are fixed to the NuFit 5.0 [7] values. The log-likelihood is modelled as a combination of Poisson distributions for the expected number of events in each bin and Gaussian distributions associated with the nuisance parameters:

$$-2 \log L(\vec{\theta}) = \min_{\vec{\epsilon}} \left\{ 2 \sum_{i,j} \left[ (N_{ij}^{\text{mod}}(\vec{\theta}; \vec{\epsilon}) - N_{ij}^{\text{dat}}) + N_{ij}^{\text{dat}} \log \left( \frac{N_{ij}^{\text{dat}}}{N_{ij}^{\text{mod}}(\vec{\theta}; \vec{\epsilon})} \right) \right] + \right.$$  

$$+ \sum_k \left( \frac{\epsilon_k - \langle \epsilon_k \rangle}{\sigma_k} \right)^2 \right\}.$$

$N_{ij}^{\text{mod}}$ and $N_{ij}^{\text{dat}}$ represent the number of reconstructed events in bin $(i,j)$ expected by the model and the number of observed events, respectively. The parameters of interest, $\vec{\theta}$, are $\theta_{23}$ and $\Delta m^2_{31}$ in the standard oscillation analysis. The rest of the oscillation parameters is kept fixed. The parameters of the model that characterise the distributions ($\vec{\epsilon}$) are composed by nuisance parameters which are related to systematic uncertainties. Some of these parameters are constrained with priors representing constraints from other experiments. Specifically:

1. Normalisations: the overall normalisation as well as the relative normalisations of the High Purity Track and Shower classes are allowed to vary with no constraints. A 20% prior uncertainty is applied to the normalisation of neutral current (NC) and $\tau$-CC events. At high energies, further approximations are included in the light propagation simulation in KM3NeT/ORCA. A 50% relative normalisation uncertainty is applied to events simulated in this regime.
2. Flux: the spectral index of the neutrino flux energy distribution, as $\phi \times E^s$, is allowed to vary from $s = 0$ with a standard deviation of 0.3. The ratio of electron neutrinos to electron antineutrinos is allowed to vary with a 7% prior uncertainty. The ratio of muon neutrinos to muon antineutrinos is allowed to vary with a 5% prior uncertainty. The ratio of muon neutrinos to electron neutrinos is allowed to vary with a 2% prior uncertainty. The ratio of vertical to horizontal neutrinos, introduced as $1 + r_{\nu\bar{\nu}} \cos \theta$, is allowed to vary from $r_{\nu\bar{\nu}} = 0$ with a standard deviation of 0.02.

3. The absolute energy scale of the detector is allowed to vary with a 9% prior uncertainty. The energy scale is related to the uncertainty on water optical properties and on the knowledge of the PMT efficiencies.

4. Results

The ORCA6 dataset has been studied to determine the neutrino oscillation parameters. Specifically, the focus in this section is on constraining the oscillation parameters $\Delta m^2_{31}$ and $\theta_{23}$.

The model is fitted to the dataset using 2-dimensional histograms on the reconstructed energy and direction, as illustrated in figure 4. Figure 5 shows the results of the fit transformed to the $L/E$ (path length over neutrino energy) ratio and normalised with respect to the "non-oscillations" hypothesis for illustration purposes.

**Figure 4**: Negative log-likelihood landscape as a function of the reconstructed cosine of the zenith angle and energy for the three classes, High Purity Tracks (left), Low Purity Tracks (middle) and Showers (right). Total negative log-likelihood is reported per class.

**Figure 5**: Ratio to non-oscillations as a function of the reconstructed path length over reconstructed neutrino energy, $L/E$, for data (black), the best-fit (blue), and NuFit (red) for the three classes: High Purity Tracks (left), Low Purity Tracks (middle), and Showers (right). Non-oscillations and NuFit hypotheses are computed taking the best-fit and fixing the oscillation parameters to the corresponding hypothesis.
The fit deviates from the expectations derived from the NuFit model by less than \( 1\sigma \). There is a preference for NO over IO: \( -2\log \left( \frac{L_{NO}}{L_{IO}} \right) = 0.9 \). The best-fit values for the parameters are \( \sin^2 \theta_{23} = 0.51^{+0.06}_{-0.07} \) and \( \Delta m^2_{31} = 2.14^{+0.25}_{-0.36} \cdot 10^{-3} \text{eV}^2 \).

The space of oscillation parameters has been scanned and profiled in terms of the negative log-likelihood, in order to provide 1-dimensional scans and a 2-dimensional contour of the sensitivity of the ORCA detector for neutrino oscillations. Figure 6 shows the 1-D scans of the profiled likelihood with 68\% and 90\% CL bands, computed by generating pseudo-experiments with the best-fit values of oscillation and nuisance parameters. The 90\% CL contour for both parameters constrained simultaneously is shown in figure 7 in comparison to other experiments.

**Figure 6**: Profiled negative log-likelihood scan of the oscillation parameters, \( \sin^2 \theta_{23} \) (left) and \( \Delta m^2_{31} \) (right). Observed limits are compatible within the 68\% CL bands.

**Figure 7**: Contour at 90\% CL of ORCA6 for the oscillation parameters \( \sin^2 \theta_{23} \) and \( \Delta m^2_{31} \) compared with other experiments. A previous result obtained with a first ORCA6 dataset is included [4].

Figure 8 shows the effect of the different systematic uncertainties. The impact is computed comparing the nominal best-fit value of \( \theta_{23} \) and \( \Delta m^2_{31} \) with the result of the fit when fixing the considered nuisance parameter shifted by \( \pm \) its post-fit uncertainty. Black dots with error bars
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represent the pulls of the parameter's best-fit. The error bars are computed as the post-fit uncertainty divided by the pre-fit uncertainty (priors). If the parameter was unconstrained, the pull is computed based on the post-fit uncertainty and the error bar is 1.

Figure 8: Impact of the different systematic parameters computed as a shift in the parameter of interest, $\Delta m^2_{31}$ (left) and $\theta_{23}$ (right) when the nuisance parameter is shifted and fixed. Systematic pulls are reported as black dots.

This test provides insights into the parameter correlations. Specifically, $\Delta m^2_{31}$ exhibits strong correlations with the systematic due to energy scale, energy spectral index, and the overall normalisation. On the other hand, correlations involving $\theta_{23}$ are generally small, although non-negligible contributions are observed from the energy scale, $\Delta m^2_{31}$, and the overall normalisation nuisance parameters.

5. Conclusions

With only 5% of its final configuration, the ORCA detector starts to contribute to the measurement of atmospheric neutrino oscillations. The best fit values for the parameters are $\sin^2 \theta_{23} = 0.51^{+0.06}_{-0.07}$ and $\Delta m^2_{31} = 2.14^{+0.36}_{-0.25} \times 10^{-3}$ eV$^2$ with a preference for NO: $-2 \log \left( \frac{L_{\text{NO}}}{L_{\text{IO}}} \right) = 0.9$. The detector deployment is progressing continuously and these measurements will gain in precision as the detector volume increases and the reconstruction and selection efficiencies are improved.

References

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