

Search for an anomalous excess of charged-current ν_e interactions without pions in the final state with the MicroBooNE experiment

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This article presents a measurement of ν_e interactions without pions in the final state using the MicroBooNE experiment and an investigation into the excess of low-energy electromagnetic events observed by the MiniBooNE collaboration. The measurement is performed in exclusive channels with $(1eNp0\pi)$ and without $(1e0p0\pi)$ visible final-state protons using 6.86×10^{20} protons on target of data collected from the Booster Neutrino Beam at Fermilab. Events are reconstructed with the Pandora pattern recognition toolkit and selected using additional topological information from the MicroBooNE liquid argon time projection chamber. Using a goodness-of-fit test the data are found to be consistent with the predicted number of events with nominal flux and interaction models with a p -value of 0.098 in the two channels combined. A model based on the low-energy excess observed in MiniBooNE is introduced to quantify the strength of a possible ν_e excess. The analysis suggests that, if an excess is present, it is not consistent with a scaling of the ν_e contribution to the flux as predicted by the signal model used in the analysis. Combined, the $1eNp0\pi$ and $1e0p0\pi$ channels do not give a conclusive indication about the tested model, but separately they both disfavor the low-energy excess model at $>90\%$ CL. The observation in the most sensitive $1eNp0\pi$ channel is below the prediction and consistent with no excess. In the less sensitive $1e0p0\pi$ channel the observation at low energy is above the prediction, while overall there is agreement over the full energy spectrum.

I. INTRODUCTION

Neutrino physics has entered an era of precision measurements of the parameters that describe three-flavor oscillations [1]. At the same time, a broad set of experimental results, collectively referred to as short-baseline anomalies [2–6], is in tension with the three-neutrino paradigm and remains without resolution. These short-baseline anomalies have often been linked to the physics signature of $\mathcal{O}(\text{eV})$ sterile neutrinos [7]. Recent long-baseline sterile-neutrino oscillation searches [8, 9], however, show tension with this interpretation, and other explanations for these anomalies may need to be considered. The observation of an excess of low-energy electromagnetic activity by the MiniBooNE experiment [5] is one example of these anomalies. Many scenarios have been suggested to explain the origin of the MiniBooNE excess of low-energy electromagnetic showers, including new physics such as sterile neutrino oscillations and decay [10, 11], dark-sector portals [12–14], heavy neutral leptons [15, 16], non-standard Higgs models [17–19], or Standard Model processes such as an enhancement of photon backgrounds [20].

The MicroBooNE experiment [21] was built to explore the nature of the low-energy excess of events observed by MiniBooNE. Operating in the same Fermilab Booster Neutrino Beamline (BNB), it is in a position to examine the nature of low-energy electromagnetic activity with the capabilities of the liquid argon time projection chamber (LArTPC) detector technology. A set of analyses,

including the one presented here, have been designed to measure both electron neutrino interactions in multiple topologies [22–24], as well as single photon events [25]. This article presents a measurement of the rate of charged current (CC) electron neutrino interactions without pions in the final state, and investigates the possibility of low-energy ν_e interactions as an explanation for the MiniBooNE observation of an anomalous excess.

Electron neutrinos that undergo CC interactions will produce a visible electron in the detector, which develops into an electromagnetic (EM) shower, along with possible hadronic activity such as protons and pions. This analysis performs a measurement of ν_e interactions with any number of protons and without visible pions in the final state ($1eXp0\pi$, $X \geq 0$). This is designed to match MiniBooNE’s single electron-like Cherenkov ring signal topology [26]. The presence of visible protons ($1eNp0\pi$, $N \geq 1$) provides additional handles for background rejection. Furthermore, the final state with no protons ($1e0p0\pi$) may be sensitive to a broad range of models [12–15, 18, 27] that predict single-shower events and go beyond the electron neutrino interpretation of the MiniBooNE anomaly. Together these motivate the choice to develop separate, orthogonal selections for events with and without protons in the final state. We focus on the use of calorimetric and topological information provided by the LArTPC technology to perform particle identification (PID) and measure electron neutrino interactions in a broad phase space. Example candidate events in these topologies from data are shown in Fig. 1. The neutrino flux and interaction systematic uncertainties associated with the selected electron neutrinos are constrained using a high-statistics inclusive measurement of CC muon

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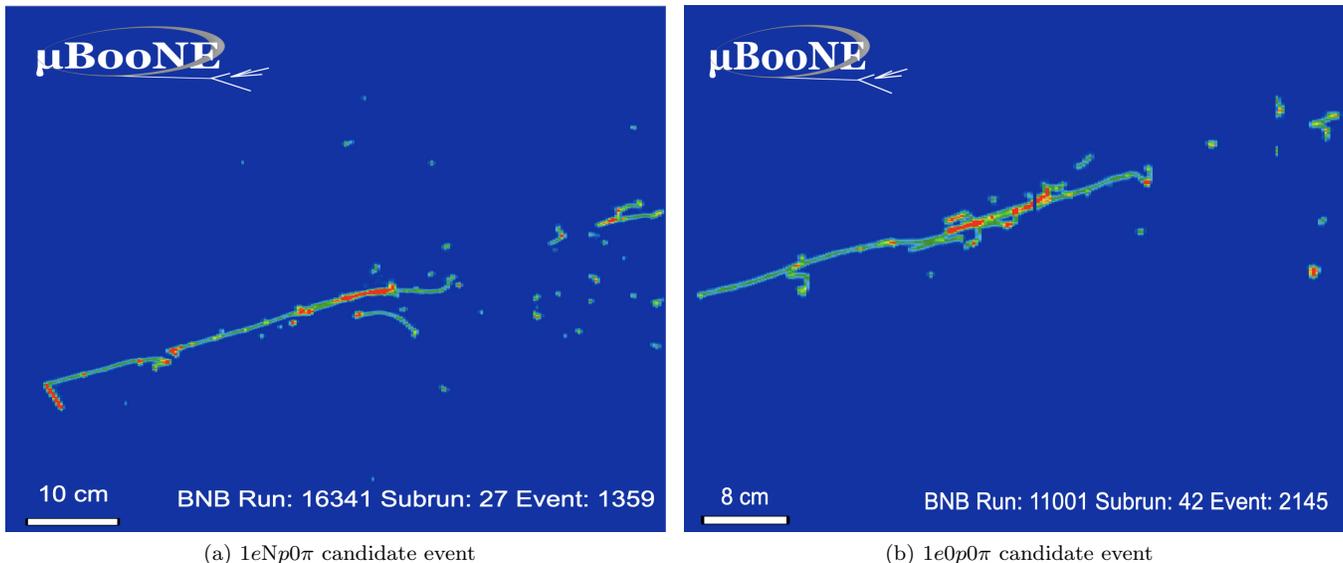


FIG. 1. Event displays of selected electron neutrino candidate data events. The horizontal axis corresponds to the wire number, which is converted into a distance based on the wire spacing, the vertical axis corresponds to the time of the recorded charge, which is converted to a distance along the TPC drift direction using the drift velocity in the TPC drift direction, and the color scale corresponds to the deposited charge. The $1eNp0\pi$ event shown (a) has a long electron shower and a short proton track attached at the vertex with a large amount of deposited energy. The $1e0p0\pi$ event shown (b) consists of a single electron shower.

neutrino interactions in the BNB. Analysis results are obtained through a series of statistical tests with the introduction of an empirical model which interprets the MiniBooNE anomaly as an enhancement of the flux of low energy electron neutrinos.

This article is organized as follows. Section II describes the neutrino beamline and MicroBooNE detector. Section III provides details of the tools used to simulate neutrino events. Section IV presents the reconstruction methods used to identify neutrino interactions. Section V presents the PID methods as well as the ν_μ and ν_e event selections. Section VI describes the blinding procedure and studies on data sidebands. Section VII details the formalism of the procedure used to reduce uncertainties based on the ν_μ observation, referred to as the ν_μ constraint. Section VIII presents the analysis results.

II. BEAMLINER AND DETECTOR OVERVIEW

This section provides a brief overview of the Booster Neutrino Beamline, the MicroBooNE detector, and the dataset used for the analysis. The MicroBooNE detector sits at a distance of 468.5 m from the BNB target, on-axis with respect to the neutrino beam. The neutrino beam begins with 8 GeV protons extracted from the Fermilab Booster synchrotron. These protons interact with a beryllium target and produce pions and kaons, which then decay to produce neutrinos. The resulting neutrino beam is composed predominantly of muon neutrinos with a small ($< 1\%$) electron neutrino component.

This electron neutrino component produced by meson decay chains in the BNB is referred to as “intrinsic ν_e ” in this article. The BNB is structured in spills, each with a duration of $1.6 \mu\text{s}$ and an intensity of up to 5×10^{12} protons, with an average repetition rate of up to 5 Hz. Additional details on the BNB are found in Ref. [28, 29].

The MicroBooNE detector [21] consists of a time projection chamber (TPC) and a photon detection system. The TPC measures 2.56 m (drift coordinate, x) \times 2.32 m (vertical, y) \times 10.36 m (beam direction, z) and contains 85 tonnes of liquid argon in its active volume. Charged particles traversing the detector ionize the argon leaving trails of ionization electrons which drift under the 273 V/cm electric field towards the anode where three planes of wires record induced currents and collect the ionization electrons. The three planes of wires, spaced 3 mm apart and oriented at 0 degrees (vertical) and at ± 60 degrees, produce three different two-dimensional views of the neutrino interaction and allow for three-dimensional reconstruction with $\mathcal{O}(\text{mm})$ spatial resolution. The low-noise TPC electronics allow for measurement of the charge with few percent resolution [30]. Combined, these features enable the MicroBooNE detector to record the final-state particles produced by neutrino interactions with the detail required to perform particle identification and accurately measure particle kinematics. The light detection system, composed of 32 photomultiplier tubes (PMTs), has a timing resolution of $\mathcal{O}(\mu\text{s})$, which allows us to select events in the BNB time window and to remove a large fraction of the cosmic-ray background. In addition, a cosmic-ray tagger (CRT) [31]

is used to reject cosmogenic interactions with precise spatial and timing information.

MicroBooNE has collected approximately 12×10^{20} protons on target (POT) of BNB data since October 2015. This analysis uses data corresponding to a total of 6.86×10^{20} POT collected between February 2016 and July 2018, which corresponds to the first three operation run periods.

III. SIMULATION AND MODELLING

This section describes the tools used in this analysis to model the neutrino flux, neutrino interactions on argon, and the detector response, as well as the low-energy excess signal. The simulation packages referenced below are used in the LArSoft software framework [32].

A. Flux model

The neutrino flux prediction from the BNB at MicroBooNE is made using the flux simulation developed by the MiniBooNE collaboration [29] and takes into account the different detector positions. The MicroBooNE and MiniBooNE experiments operated simultaneously between 2015 and 2019. No significant variations in the flux were observed by the MiniBooNE collaboration during this time [33]. The flux prediction is therefore assumed to be invariant over time.

There are two primary sources of uncertainty in the beam flux modeling. The first is hadron production cross sections, which includes the production of π^\pm , K^\pm , and K_L^0 particles in the beamline. The second is related to the beamline, in particular the modeling of the horn configuration and current, as well as of secondary interactions on the beryllium target and aluminum horn. Flux uncertainties in this analysis are treated analogously to those in MicroBooNE cross section measurements [34–39] and follow the implementation of the MiniBooNE collaboration as described in Ref. [29] including the improvements described in Ref. [40]. Flux uncertainties have a 5–10% impact on the event rate after selection and, for neutrinos with energy below 0.8 GeV, are dominated by hadron production uncertainties.

B. Neutrino interaction model

MicroBooNE relies on the GENIE [41] event generator to simulate neutrino interactions in the detector and model the outgoing final-state particles produced. This analysis uses GENIE v3.0.6 G18_10a_02_11a, which incorporates theoretical models and experimental results relevant for 0.1–1 GeV neutrino interactions. A tune of this model was developed by fitting parameters of particular importance to the modeling of sub-GeV CC interactions to external data [42].

Modeling uncertainties on neutrino interactions in GENIE are obtained in three different ways. For model parameters that are estimated using the tune, we use the parameter uncertainty from the fit. For all other parameters in the model we use uncertainties as provided by the GENIE collaboration. Finally, for parameters for which an uncertainty was not provided, we estimate the uncertainty in other ways, such as by choosing the full range between different available models that cover the world data. The treatment of systematic uncertainties in the neutrino interaction model is further detailed in Ref. [42], including an inconsistency in the treatment of the final state interaction uncertainty which has a negligible impact on the analysis. The effect of cross section uncertainties on the predicted neutrino event rate after selection is about 20%, and constitutes the leading source of systematic uncertainty for the analysis.

C. Detector model

MicroBooNE’s detector response is modeled with multiple simulation tools. Geant4 [43] v10_3_03c is used to simulate the propagation of particles through the detector. The propagation of light and charge in the detector is done within LArSoft.

The MicroBooNE TPC readout electronics and wire response are determined using a simulation of the induced charge from drifting electrons [30, 44]. The production of scintillation light is simulated through a voxelized look-up library created from a detailed Geant4 simulation to model photon propagation. Several detector simulation components are implemented using a data-driven approach. These involve effects that lead to a non-uniform detector response in space and time. Electric field distortions due to space-charge buildup in the active volume are accounted for through MicroBooNE’s data-driven electric field maps [45, 46]. Non-uniformities across the detector due to electron lifetime or wire response are simulated in a time-dependent way when appropriate. Ion recombination is simulated using a modified box model [47].

As MicroBooNE is a surface detector, cosmic rays are the largest background to neutrino interactions. A data-driven method is used to eliminate the need to simulate the high rate of cosmic rays passing through the detector as well as intrinsic noise in the TPC and PMT electronics. This starts with a dedicated data stream which is collected in periods when there is no beam and provides a sample of detector activity from both cosmic rays and electronics noise. Then, to form the beam simulation, TPC and PMT waveforms from simulated neutrino interactions are merged with this beam-off data-stream, ensuring faithful modelling of cosmic-ray backgrounds and noise. We use this approach for the simulation of neutrino interactions taking place both inside and outside the TPC fiducial volume; the latter are referred to as “dirt” background events in this analysis and include in-

interactions in the LAr outside the TPC fiducial and in the walls of the cryostat as well as the rock around the detector cavern.

Detector response systematic uncertainties include the propagation of final-state particles as well as the formation of light and charge signals. Uncertainties associated with the TPC dominate over the ones associated with the light collection system, with a total uncertainty of 10–15% in the ν_e event rate. Mis-modeling of the wire response, electric field map, and ion recombination each contribute at a smaller level but in similar magnitude. The treatment of wire response systematics are discussed in [48]. The impact of final state particle propagation is assessed by varying charged pion and proton reinteraction cross sections available from external data, using the `Geant4Reweight` [49] framework. This leads to $\mathcal{O}(1\%)$ uncertainties on the event rate. Systematic uncertainties associated with light production impact only the first stages of the analysis related to cosmic-ray rejection and lead to uncertainties of 3–5% on the event rate. Uncertainties due to the limited sample size are also included in the analysis and in terms of ν_e event rate they vary from $\mathcal{O}(1\%)$ for the most populated bins to $\mathcal{O}(100\%)$ for the bins with very small prediction at high energy.

D. Unfolded median MiniBooNE ν_e excess model

This analysis searches for an excess of electron neutrino events over the predicted intrinsic interaction rate. To benchmark the analysis performance and calculate sensitivity to potential new physics, we adopt a model constructed using the MiniBooNE dataset to obtain a prediction of a ν_e -like excess in the MicroBooNE detector.

To construct the model, the background-subtracted excess of data events from MiniBooNE’s 2018 result [5] is unfolded using MiniBooNE’s electron neutrino energy reconstruction smearing matrix, constructed with the NUANCE [50] neutrino interaction simulation using a CC quasi-elastic energy definition, and accounting for MiniBooNE’s energy smearing and selection efficiency. This predicts the rate of electron neutrinos as a function of true neutrino energy above 200 MeV. The ratio between the predicted rate from MiniBooNE and that of the intrinsic electron neutrino component in MiniBooNE’s simulation is used to obtain an energy-dependent flux scaling factor for the excess under the electron neutrino hypothesis. These energy-dependent weights are applied to the rate of intrinsic electron neutrino events predicted by MicroBooNE’s flux and cross section simulation to obtain a prediction for the MiniBooNE ν_e -like excess in the MicroBooNE detector. Uncertainties from the MiniBooNE measurement are not propagated in our signal prediction as an accurate determination of correlations with MiniBooNE uncertainties is beyond the scope of this work and will require a joint analysis of the two experiments. We will refer to this model in the article as the eLEE

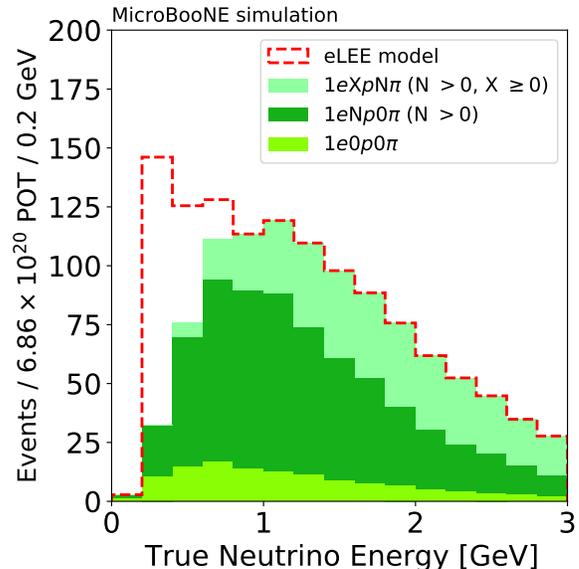


FIG. 2. Predicted electron neutrino events broken down by true number of visible protons and pions. The final states selected in this analysis ($1eNp0\pi$ and $1e0p0\pi$) are shown along with all other ν_e interactions ($1eXpN\pi$). Pion categorization refers to both charged and neutral pions.

model. Figure 2 shows the truth-level intrinsic ν_e spectrum, broken into final state particles, and the additional contribution of the prediction of eLEE model events. In this plot, and throughout this work, protons and charged pions are considered “visible” and counted if their true kinetic energy is above 40 MeV. A scaling factor μ is used to vary the normalization of the excess component of the flux. Systematic uncertainties on eLEE signal events are applied analogously to those for intrinsic ν_e interactions, consistent with their implementation as an enhancement of the intrinsic ν_e flux. Since it is constructed using an unfolding procedure based on the neutrino energy, this model is only used to predict the event rate as a function of reconstructed neutrino energy and not for predictions in other kinematic variables.

Although it does not fully characterize the MiniBooNE excess, this empirical model provides a benchmark that allows the analysis to quantitatively relate to the ν_e interpretation of the MiniBooNE excess and to provide a reproducible reference for further interpretations. As mentioned in Sec. I however, different theoretical models, such as sterile neutrino oscillations or more exotic phenomena, can lead to different predictions in MicroBooNE. These are not directly explored here and are left for future work.

IV. NEUTRINO EVENT RECONSTRUCTION

This section describes the methods used to reconstruct events, separate cosmic-rays from neutrino interactions,

and calculate neutrino energy.

A. Event reconstruction

Event reconstruction in the MicroBooNE TPC starts with the processing of the electronic signal waveforms recorded on wires, which includes noise suppression [51] and signal processing [30, 44]. Gaussian-shaped pulses on the waveforms, called “hits”, are then identified, and the associated wire number, time, and integrated charge are inputs to later reconstruction steps.

This analysis uses the Pandora [52] event-reconstruction. Pandora is a multi-algorithm pattern recognition toolkit that performs particle tracking for LArTPC detectors. It has been extensively developed within the MicroBooNE collaboration and used for numerous published results [34–39, 53–55]. It starts with the two-dimensional hit coordinates from each plane and outputs three-dimensional particles categorized in terms of their hierarchy in the neutrino interaction. In this hierarchy primary particles are directly produced in the neutrino interaction, while those from their decay or interaction with the argon are secondaries. Particles are classified as shower-like (electrons and photons) or track-like (muons, charged pions, and protons) using a score ranging from 0 to 1. The Pandora pattern recognition is further complemented by analysis-specific tools that enhance the PID and track-shower separation capabilities with an emphasis on ensuring powerful muon/proton and electron/photon separation for particles from low energy neutrino interactions (see Sec. V A).

TPC detector calibrations implement position- and time-dependent corrections that provide a uniform detector response in addition to absolute gain calibrations necessary for calorimetric energy measurements and PID. These include a calibration of the position-dependent electric field [45] using MicroBooNE’s UV-laser as well as calibrations of the electron lifetime, wire-response, and absolute charge. MicroBooNE’s overall calibration strategy is described in Ref. [56] and relies on through-going and stopping cosmic muons.

B. Cosmic-ray rejection

The near-surface location and the low neutrino interaction rate in the detector lead to significant cosmic-ray contamination. For each beam spill, $\mathcal{O}(10)$ cosmic rays cross the detector in the 2.3 ms TPC drift window. Conversely, approximately one in every 1000 spills lead to a neutrino interaction in the active volume for a ν -to-cosmic ratio of 10^{-4} . Scintillation light is used to suppress cosmic-ray backgrounds, first as part of an online trigger selection and subsequently through an offline analysis filter. A requirement to observe prompt scintillation light in coincidence with the beam rejects 98% of recorded beam-spills while accepting more than 99%

for ν_e interactions with neutrino energy greater than 200 MeV.

At this stage in the analysis, selected events are still dominated by cosmic-ray interactions which occur in time with the 1.6 μ s BNB spill. Through-going or out-of-time TPC interactions are rejected as obvious cosmic rays and removed from further analysis. The three-dimensional charge-pattern of the remaining interactions identified in the TPC is compared to the pattern of scintillation light collected on the detector’s PMTs. Compatibility between the absolute charge and light, as well as their relative position in the TPC, is required. Including these tools in the selection leads to an additional suppression of cosmic-ray interactions by a factor of ten, with an integrated efficiency of 83% for both ν_e and ν_μ CC interactions.

The CRT [31] provides an additional tool for cosmic rejection. CRT information is available only for data taken after December 2017, when it was integrated in MicroBooNE’s analysis chain. Its usage in this analysis is limited to the ν_μ selection (Sec. V B).

C. Energy reconstruction

The MicroBooNE LArTPC can detect particles with a threshold of few to tens of MeV and measure the energy deposited in the neutrino interaction with high precision. Energy reconstruction is performed calorimetrically for electromagnetic (EM) showers and based on measurements of particle range for track-like particles. Selections in this analysis require particle containment in the detector (see Sec. V). Range-based track energy measurements deliver very good energy resolution, which is estimated from simulation to be 3% for muons and $< 2\%$ ($< 9\%$) for protons with kinetic energy > 100 MeV (> 40 MeV). The EM shower energy is measured by integrating the deposited energy ($E_{\text{calorimetric}}$) and relying on simulation of electron showers to obtain a correction factor which accounts for inefficiencies in collecting the full energy deposited [57]. This leads to a reconstructed energy definition of $E_{\text{corrected}} = E_{\text{calorimetric}}/0.83$. The electron energy resolution is $\leq 12\%$ and is dominated by the charge clustering inefficiencies discussed above. The reconstructed neutrino energy for ν_e and ν_μ interaction candidates is calculated using:

$$E_{\text{reco}}^{\nu_e} = E_{\text{corrected}}^{\text{electron}} + \sum_{\text{tracks}} E_{\text{range}}^{\text{proton}} \quad (1a)$$

$$E_{\text{reco}}^{\nu_\mu} = E_{\text{range}}^{\text{muon}} + \sum_{\text{other tracks}} E_{\text{range}}^{\text{proton}} + 0.105 \text{ GeV} \quad (1b)$$

where 0.105 GeV is the muon mass. In the energy definition we assume that all tracks other than the selected muon are protons, which matches the ν_e selection without pions, but represents an approximation for the inclusive ν_μ selection. This definition achieves 15% energy

resolution for both selected ν_e and ν_μ events in the low energy region primarily targeted by this analysis. For ν_e events this definition measures the energy deposited by charged final state particles above threshold and provides an accurate estimate, with an average bias at the percent level; when compared to the true neutrino energy, however, it typically underestimates by 16% (9%) for selected $1e0p0\pi$ ($1eNp0\pi$) events. More details on PID are described in Sec. V.

V. NEUTRINO EVENT SELECTIONS

Neutrino candidate events are initially identified using the reconstruction methods described in Sec. IV. The following section presents a description of several of the PID tools developed for this analysis as well as the ν_μ and ν_e selections in which they are used.

A. Particle identification

The primary PID tasks required for this analysis are the separation of highly ionizing proton tracks from minimally ionizing muons and pions as well as the separation of photon and electron electromagnetic showers. To distinguish stopping muons from protons we leverage the difference in the energy loss profile at the Bragg peak through a measurement of the energy loss per unit length (dE/dx) versus particle residual range. A probability density function for simulated protons and muons is used to construct a likelihood function that combines the measured dE/dx at each point along a particle's trajectory from the calorimetric information on all three planes [58]. This tool provides a 90% relative efficiency for proton selection with a 5% mis-identification rate. Track PID is used to identify muon candidates produced by ν_μ CC interactions, isolate protons, and remove pion candidates.

Two key features are used to achieve electron-photon separation: the calorimetric measurement of dE/dx at the start of the shower and the displacement of the electromagnetic shower's start position from the primary vertex in neutrino interactions with hadronic activity. To evaluate dE/dx , reconstructed showers are fit using a Kalman filter [59] based procedure to identify the main shower trunk and reject hits that are transversely or longitudinally displaced. Values of dE/dx measured in the first few centimeters of the electromagnetic shower, before it starts to cascade, are used to compute a median dE/dx characteristic of the shower's energy loss [60]. Information from all three wire planes is used to optimize the ability to perform electron-photon separation independently of particle orientation. Multiple ranges at the shower start point are used to evaluate dE/dx to account for the potential impact of protons at the vertex and early branching of the electromagnetic shower and provide additional separation power. The dE/dx variable is shown in Fig. 3. Good separation between elec-

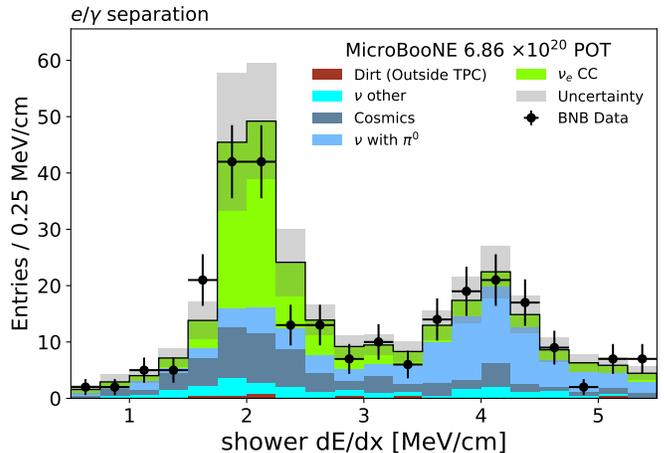


FIG. 3. Energy deposited per unit length (dE/dx) for electron-photon separation. The figure shows dE/dx measured in the $[0,4]$ cm range from the shower start point for a combination of events with and without protons. Data from the signal region ($E_\nu < 0.65$ GeV) is excluded from this validation plot. The contributions to the stacked histogram are comprised of charged-current intrinsic ν_e interactions with any number of final state hadrons in green, ν_μ and neutral-current ν_e interactions that produce one or more π^0 s in the final-state in light blue, and all other ν interactions in cyan. Dirt backgrounds are in red, and cosmic backgrounds in grayish-blue. This categorization is used in all ν_e selection figures.

tron and photon showers is observed and contributes to the π^0 background rejection achieved by this analysis. In this and other data/simulation comparison plots shown in the article data points are shown with associated statistical uncertainty, computed as \sqrt{N} , while systematic uncertainties on the prediction are shown as a shaded gray band.

B. ν_μ measurement

The vast majority of neutrinos reaching the MicroBooNE detector are muon neutrinos. They come from the same flux of parent hadrons and interact on the same target argon in the detector as the electron neutrinos. This makes the measurement of high-statistics ν_μ interactions a valuable handle with which to validate and constrain intrinsic ν_e flux and cross section systematic uncertainties. This is done with a ν_μ CC inclusive selection that allows any number of final state hadrons and prioritizes performance at low energy. A muon neutrino candidate is identified by the presence of a muon candidate inside the TPC fiducial volume. The muon is required to be contained, which preserves good efficiency for low-energy ν_μ interactions, while suppressing cosmic-ray muon backgrounds. Cosmic rays are the primary background for ν_μ CC events, and an additional 64%

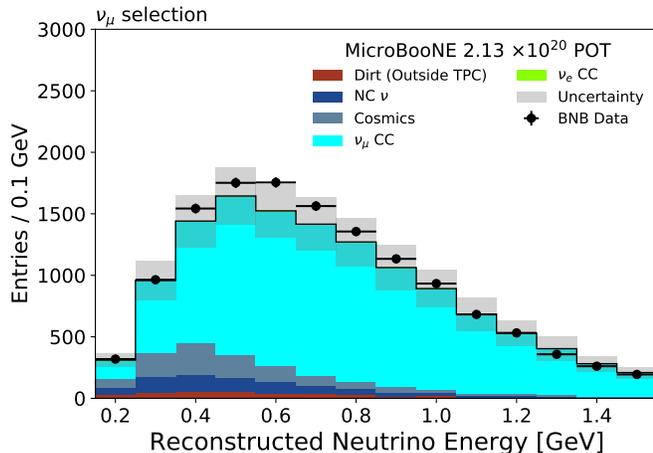


FIG. 4. Selected muon neutrino events compared to simulation split by true interaction categories. The predicted stacked distribution is comprised of charged-current ν_μ interactions in cyan, neutral-current backgrounds in dark blue, ν_e interactions in green, and cosmic backgrounds in grayish-blue. The shaded band is the systematic uncertainty.

of these are removed using the CRT. The ν_μ constraint only uses 2.13×10^{20} POT of data collected after December 2017, when the CRT was available. Events surviving these selection requirements are required to have a track PID with a muon-like value. Consistency is required between two independent measurements of muon energy: the range-based energy estimation and one based on multiple Coulomb scattering as described in Ref. [61]. The reconstructed neutrino energy distribution for the final ν_μ selection is shown in Fig. 4. The data sample contains 13,346 events with a predicted ν_μ CC purity of 77%. The main backgrounds are cosmic ray and neutral-current neutrino interactions. Data and simulation are found to agree within systematic uncertainties in reconstructed neutrino energy as well as in other muon neutrino kinematic variables. These include the muon energy and angle with respect to the beam, and were tested quantitatively accounting for all uncertainties and their correlations.

C. ν_e selections

Electron neutrinos are measured with two separate selections targeting events with and without visible protons. These are referred to as the $1eNp0\pi$ and $1e0p0\pi$ selections where $N \geq 1$. In simulation, we define a proton as visible if it has a kinetic energy of at least 40 MeV. Combined, these span the signal definition of electron neutrinos measured by the MiniBooNE experiment: events with a single electron, any number of protons, and no pions.

The analysis targets contained ν_e interactions occurring in the fiducial volume, defined by a boundary

of 10 cm in the drift coordinate, 15 cm in the vertical, and 10 and 50 cm from the front and end of the TPC respectively. The selections rely on a common pre-selection which identifies an event as ν_e -like. An event is defined as ν_e -like if there is a contained reconstructed electromagnetic shower with at least 70 MeV of deposited energy. The reconstructed energy requirement removes Michel electrons from cosmic- or neutrino-induced muons. Events are then further classified depending on the presence or absence of proton candidates.

The $1eNp0\pi$ and $1e0p0\pi$ selection definitions split after pre-selection, but the strategy and inputs used for the following steps are the same for both. Events are classified based on topological and calorimetric information such as the track PID score and dE/dx as described earlier. Additional handles are used to separate ν_e events from events with a π^0 . These are the distance between the neutrino interaction vertex and the start point of the shower, known as the conversion distance, and a second shower search. This analysis does not use kinematic quantities in the selection criteria to limit the model dependence of the results. A set of selection requirements called the “loose” selection is defined using these variables to remove large portions of the backgrounds for higher statistics data-simulation comparisons in more ν_e -like regions. Next, these variables are used to train boosted decision trees (BDTs) for the two channels used in the analysis.

The main backgrounds for the ν_e selections are cosmic-rays, neutrino interactions with π^0 production, and neutrino interactions (referred to as “ ν other”) that produce charged pions or muons that eventually produce a Michel electron that is mis-identified as an electron produced by a ν_e interaction. After the full selection dirt events outside the TPC fiducial volume are a negligible contribution.

1. $1eNp0\pi$ selection

The $1eNp0\pi$ channel is most sensitive to the eLEE model as it is able to use tracks associated with the vertex in addition to the shower to select electron neutrino events and reject backgrounds. In this selection, two BDTs are trained with XGBoost [62] to separate signal from background: one targets removal of backgrounds that contain a π^0 , and the other backgrounds without π^0 s. Samples of ν_e events simulated with true neutrino energy below 0.8 GeV are used to define the signal when training the BDTs. Simulated samples with ν_μ events from a variety of true interaction categories are used to train the BDT to identify backgrounds. Sixteen topological and calorimetric variables are used to distinguish signal from background in these BDTs. The most important of these are the shower conversion distance, which separates ν_e from π^0 events, and the number of distinct branches in the shower, which separates mis-reconstructed ν_μ interactions from ν_e -induced showers. The longest track-like particle in the interaction is re-

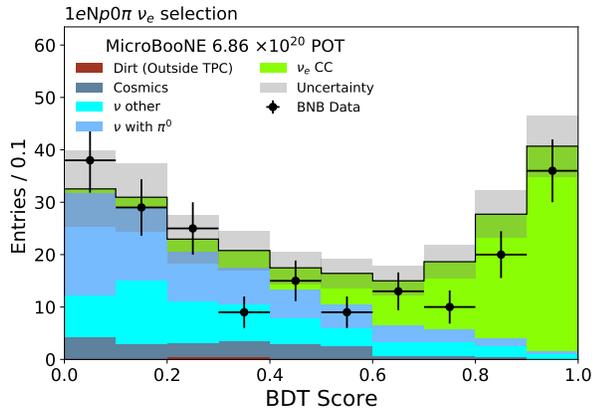


FIG. 5. Response of the $1eNp0\pi$ selection BDT designed to reject events with π^0 s. Background events are predicted to peak at low BDT scores and electron neutrinos at high BDT scores. Events with BDT score above 0.67 are retained as part of the final selection. Gray bands denote the systematic uncertainty on the prediction.

quired to be proton-like which further helps to suppress cosmic-ray backgrounds, ν_μ backgrounds, and ν_e interactions with final-state charged pions. At pre-selection the purity of the $1eNp0\pi$ selection is expected to be at the percent level. After the full selection is applied the ν_e purity is expected to be 80% with an efficiency of 15% for true $1eNp0\pi$ events defined based on the 40 MeV proton energy threshold. The response of the BDT targeting events with π^0 's is shown in Fig. 5 for the full data-set after the loose selection. The selected sample is obtained by rejecting events with BDT score less than 0.67 and 0.70 for the π^0 and non- π^0 BDTs respectively. Relative to pre-selection, cosmic background events are reduced by 99.98% and background events with π^0 s are reduced by 99.93%. The predicted composition of the selected $1eNp0\pi$ sample is shown in Table I. The selected ν_e CC events are predicted to be 95% true $1eNp0\pi$ events, with a $\sim 5\%$ contamination of events with pions.

TABLE I. Predicted composition of the $1eNp0\pi$ selected events with unconstrained systematic uncertainties in the reconstructed neutrino energy range 0.01–2.39 GeV for 6.86×10^{20} POT.

1eNp0π Selection	
True Category	Predicted Events
ν_e CC 0p0π	0.4 ± 0.1
ν_e CC Np0π	71.7 ± 10.6
ν_e CC XpNπ	3.3 ± 0.9
ν_e CC total	75.4 ± 11.0
ν with π^0	5.1 ± 1.4
ν other	5.5 ± 1.1
Cosmic-rays	0.8 ± 0.5
Total	86.8 ± 11.5

2. $1e0p0\pi$ selection

The $1e0p0\pi$ topology is sensitive to ν_e events in the eLEE model, as well as potentially to single-electron events from a broader range of models. In addition, it complements the $1eNp0\pi$ selection by mitigating migration effects that may arise from mis-reconstruction or mis-modeling of the multiplicity and kinematics of protons produced by neutrino interactions.

A single BDT is trained to select true $1e0p0\pi$ events and true $1eNp0\pi$ events in which protons are not reconstructed. The methods used are the same as those for the $1eNp0\pi$ selection described in Sec. V C 1, except that only a single BDT is used to reject backgrounds. The BDT leverages 28 topological and calorimetric variables, the most important of which are the measurements of dE/dx which separate electrons from π^0 s. The BDT response is shown in Fig. 6 for the full data set after applying the loose selection. The final selection is made by requiring events have a BDT score greater than 0.72.

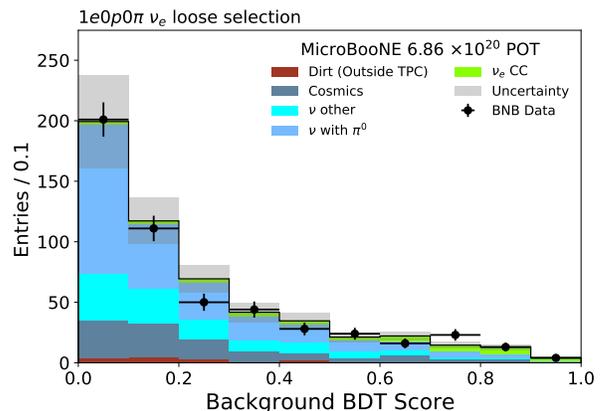


FIG. 6. $1e0p0\pi$ selection BDT response. Background events are predicted to peak at low BDT scores and electron neutrino events at high BDT scores. In the final selection, events with BDT score above 0.72 are retained. Gray bands denote the systematic uncertainty on the prediction.

After pre-selection the ν_e purity is estimated to be at the percent level. After the full selection is applied the ν_e purity is expected to be 43% with an efficiency of 9% for true $1e0p0\pi$ events. The selected ν_e events are predicted to be 70% true $1e0p0\pi$ events and 30% true $1eNp0\pi$ events. Relative to pre-selection cosmic background events are reduced by 99.8% and the backgrounds from events with π^0 s are reduced by 99.7%. Even with this level of π^0 background suppression, the overall π^0 contribution to the predicted event rate is, at low energies, comparable to that of electron neutrinos. This is due to the relatively low rate of $1e0p0\pi$ interactions as well as residual reconstruction limitations. The predicted number of events after the BDT selection is shown in Table II.

TABLE II. Predicted composition of the $1e0p0\pi$ selected events with unconstrained systematic uncertainties in the reconstructed neutrino energy range 0.01–2.39 GeV for 6.86×10^{20} POT.

1e0p0 π Selection	
True Category	Predicted Events
ν_e CC 0p0 π	8.7 ± 3.0
ν_e CC Np0 π	3.8 ± 0.7
ν_e CC XpN π	0.3 ± 0.1
ν_e CC total	12.8 ± 3.4
ν with π^0	8.6 ± 1.9
ν other	3.1 ± 1.1
Cosmic-rays	5.7 ± 1.5
Total	30.1 ± 4.3

VI. SIDEBANDS AND BLIND-ANALYSIS STRATEGY

This measurement of the ν_e rate in the BNB and the corresponding exploration of the ν_e nature of the MiniBooNE excess was designed as a blind analysis, without access to the ν_e component of the BNB flux. This choice minimizes the risk of bias but also requires careful validation. The flux, cross section, and detector models used in the ν_e selections are validated using numerous data sidebands, which include samples dominated by ν_μ and π^0 backgrounds, as well as the NuMI [63] neutrino beam data. In addition, a small amount of BNB data, less than 10% of the total data set, was fully open during analysis development. Each of these sidebands are described in the following sections, followed by a description of the unblinding procedure. All sidebands are orthogonal with respect to the signal selection.

A. Background-enriched sidebands

Multiple sideband samples were developed to validate the background modeling. Neutral pions are particularly useful as they are both a well-understood standard candle to validate the shower energy reconstruction and an important background for the ν_e measurement [57]. The area-normalized data to simulation comparison of the reconstructed invariant mass, $M_{\gamma\gamma}$, from $\pi^0 \rightarrow \gamma\gamma$ decays in a high-statistics π^0 sample is shown in Fig. 7. It demonstrates good reconstruction performance for EM showers and well modeled energy-scale calibration. An additional sideband dominated by neutrino interactions in which a π^0 is produced was developed by applying the ν_e selection but requiring that there be at least two reconstructed showers instead of one. This sample is π^0 rich and its similarity to the ν_e selections makes this sideband ideal for validating the predicted π^0 background to the ν_e measurement. Overall, the prediction was found to be consistent with the observation in π^0 rich selections, with a trend showing more predicted π^0 events at higher energies compared to the observation. This trend is ac-

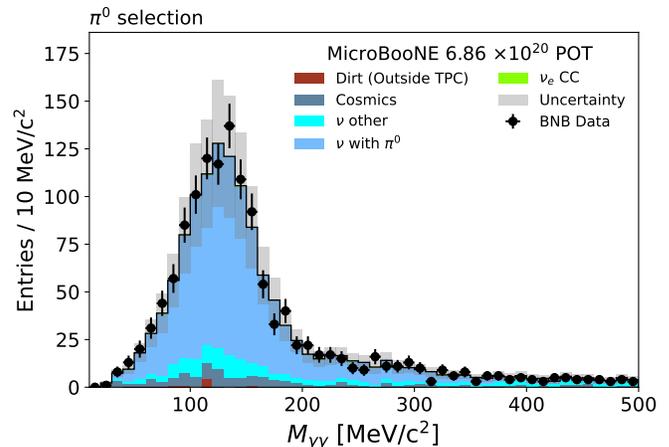


FIG. 7. Area-normalized comparison of data and simulation for the di-photon mass from π^0 candidates. The data and simulation peaks agree to within 1%, and fall within 5% of the accepted π^0 mass of $135 \text{ MeV}/c^2$, demonstrating good energy-scale calibration for EM showers.

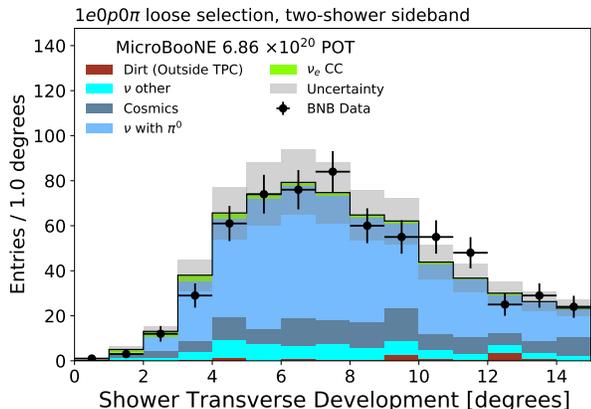
counted for by the $\mathcal{O}(20\%)$ systematic uncertainty associated with pion production in the neutrino interaction model. All inputs to the selection BDTs were checked in this sideband at each selection stage. As an example, Fig. 8a shows events in this sideband that would pass the $1e0p0\pi$ loose selection but have more than one shower. The variable plotted is the shower transverse development angle which parameterizes the shower’s extension into the plane orthogonal to its principal axis.

Sidebands addressing all known major sources of background events were obtained by inverting the selection requirement on the BDT scores used by the electron neutrino selections. The result is the definition of separate sidebands for the $1eNp0\pi$ and $1e0p0\pi$ selections. The corresponding datasets were opened progressively: first a “low BDT” sideband, and then an “intermediate BDT” sideband; all input variables were checked in these sidebands. As an example, Fig. 8b shows the distribution of the conversion distance for the reconstructed shower at pre-selection stage for the $1eNp0\pi$ low-BDT sideband. Events with a distance greater than 6 cm between the vertex and the shower are rejected by the loose selection, and the variable is further provided as input to the BDT.

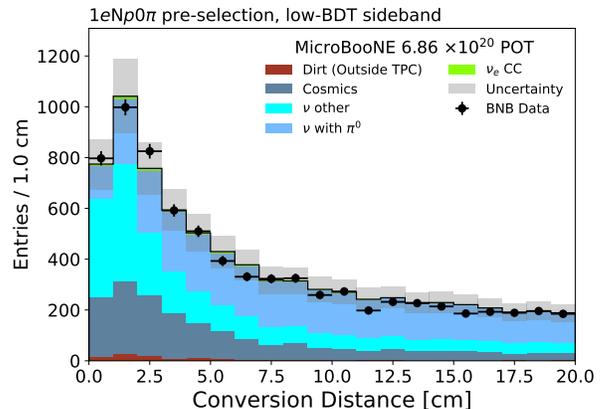
Consistency between data and simulation is assessed through quantitative goodness-of-fit tests which show excellent agreement in all sidebands and selection stages, validating the background modeling with high statistics samples.

B. NuMI beam data

Studying ν_e interactions in data is crucial in order to validate the analysis selection performance and the ν_e modeling. Measurements of ν_e interactions on argon



(a) The transverse development angle for events that would pass the $1e0p0\pi$ loose selection but have more than one shower. Most of the events contain π^0 s, and good agreement between data and simulation indicates that this background is well modeled.



(b) The shower conversion distance at pre-selection in the $1eNp0\pi$ low-BDT sideband. Background events with π^0 s are predicted to typically have a longer conversion distance than those without π^0 s.

FIG. 8. Example distributions of BDT input variables in sideband regions.

using the NuMI beamline [63] have been performed by the ArgoNeuT [64] collaboration as well as by MicroBooNE [34, 35]. This makes the NuMI beam data collected at MicroBooNE a well understood sample which is particularly valuable for this validation. Electron neutrinos from NuMI are produced almost entirely from the decay of unfocused kaons at the target, unlike those produced in the BNB which come approximately equally from pion and kaon decays. They reach the MicroBooNE detector at about 27 degrees off the TPC axis and share a similar energy range and peak, around 1 GeV, with the electron neutrinos intrinsic to the BNB, but with a narrower distribution. Results from applying the $1eNp0\pi$ and $1e0p0\pi$ selections to NuMI data from MicroBooNE's first year of operations are shown in Fig. 9. Both the $1eNp0\pi$ and $1e0p0\pi$ channels are predicted to have a $\nu_e + \bar{\nu}_e$ purity of 87%. The relatively high $\nu_e + \bar{\nu}_e$ content of the NuMI beam, 5% relative to 0.5% in the BNB, contributes to low predicted background levels in NuMI compared to those predicted in the BNB and makes the NuMI beam insensitive to new electron neutrino signatures in this analysis. The comparable ν_e and $\bar{\nu}_e$ contributions to the NuMI flux, combined with the smaller rate of final-state protons in $\bar{\nu}_e$ interactions, leads to a $\sim 40\%$ $\bar{\nu}_e$ component in the $1e0p0\pi$ prediction. In the $1e0p0\pi$ channel we observe 16 events with 16.9 predicted, and 54 with 53 predicted in the $1eNp0\pi$ channel. The good level of agreement between observed and predicted ν_e events on a beamline that has been used for multiple ν_e cross section measurements on argon provides a strong validation of the selection's ability to identify electron neutrinos in data.

C. Progressive data opening for BNB events

A small dataset was fully open during the development of this analysis: 4×10^{19} POT from the first period of data taking and 1×10^{19} POT from the third period of data taking. Together these correspond to less than 10% of the full 6.86×10^{20} POT used for the results presented in this article. This small open dataset was first used to develop and test the MicroBooNE LArTPC event reconstruction. It was also used to validate agreement between data and simulation during analysis development, where all inputs to the BDTs and other variables of physics interest were cross-checked at each selection stage. After the selections were frozen and validated in sideband data the analysis was tested with several fake data sets. These were created with and without an injected electron-like signal as well as with simulation modifications. Results with the frozen analysis on these fake data sets were found to be consistent with the true injected signal. Unblinding was performed in stages moving progressively from background-enriched sidebands and high-energy ν_e regions towards the low-energy region in the BNB where the eLEE signal is predicted according to the model described in III D. This was done in three energy regions. The high-energy region was defined as $E_{\text{reco}}^{\nu_e} > 0.85$ (0.90) GeV for the $1eNp0\pi$ ($1e0p0\pi$) selection, the medium-energy region as $E_{\text{reco}}^{\nu_e} > 0.65$ GeV for both selections, and the low-energy region as $E_{\text{reco}}^{\nu_e} > 0.15$ GeV for both selections. Selection criteria, including BDTs, were frozen before opening the first high-energy ν_e sideband. Following the opening of the medium and high energy regions, goodness-of-fit p -values of 0.277, 0.206, and 0.216 were measured respectively for $1eNp0\pi$ events, $1e0p0\pi$ events, and the two combined. These are obtained comparing the observed data to the prediction from simulation after applying the

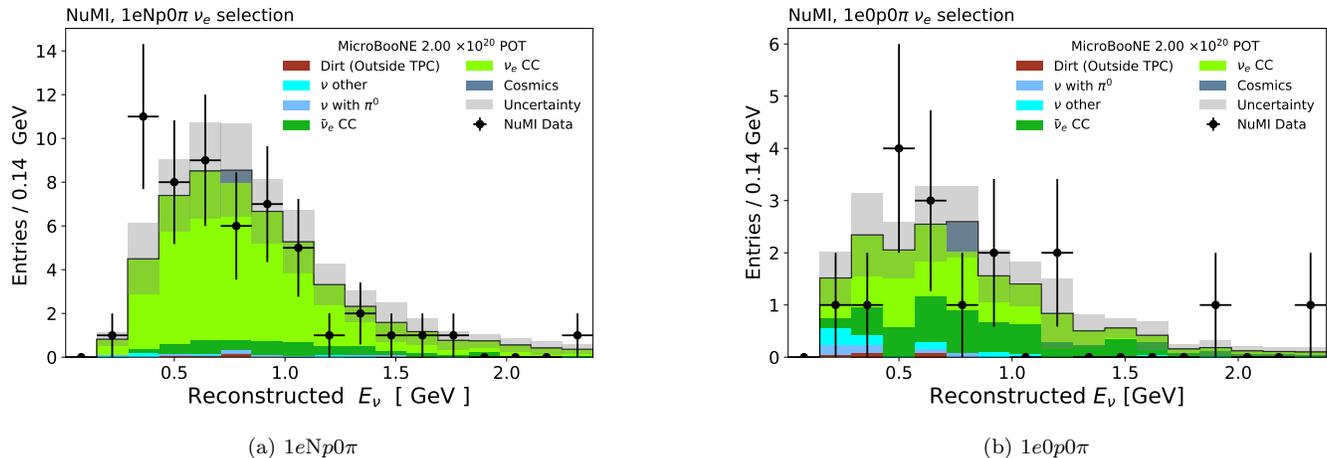


FIG. 9. Reconstructed neutrino energy for events in $1eNp0\pi$ (a) and $1e0p0\pi$ (b) selections from the NuMI beam. In the $1e0p0\pi$ sample both the relatively large $\bar{\nu}_e$ content, due to similar ν_e and $\bar{\nu}_e$ fluxes, as well as the low non- ν_e background content, due to the relatively large $\nu_e + \bar{\nu}_e$ fraction in the beam, are distinctive features in NuMI that make the spectrum different from the analogous predictions in BNB.

ν_μ constraint procedure described in the next section. The level of consistency with the underlying prediction supported the decision to move forward with unblinding the full energy range.

VII. SYSTEMATICS AND APPLICATION OF THE ν_μ CONSTRAINT

We separate the sources of systematic uncertainty into three main categories: flux, cross section, and detector response uncertainties. Details about variations for each of these categories are presented in Sec. III. Uncertainties associated with the statistics of the simulation samples used in the analysis are also included. Uncertainties related to the flux, particle propagation, and, partially, cross section are assessed through multi-universe simulations which are generated by varying the underlying model parameters within their range of uncertainty. Detector response and several cross section model uncertainties are assessed through single variations of the underlying simulation model, referred to as uni-sims. Systematic uncertainties are included in the analysis through the formalism of the covariance matrix:

$$C^{\text{Syst}} = C^{\text{Flux}} + C^{\text{XSec}} + C^{\text{Detector}} + C^{\text{MCStat}} \quad (2)$$

$$C_{ij} = \frac{1}{N} \sum_{k=1}^N \left(n_i^k - n_i^{\text{CV}} \right) \left(n_j^k - n_j^{\text{CV}} \right), \quad (3)$$

where C indicates a covariance matrix, i, j are indices over histogram bins, n^{CV} is the nominal (central value) bin content, n_i^k is the content of the i th bin in the alternative universe k , and N is the total number of alternative universes. Uni-sim variations are symmetrized in the covariance matrix approach adopted in the analysis.

Uncertainties are constrained by leveraging the correlations between ν_μ and ν_e events, which share common flux parentage in their decay chain in the beamline and significant overlap in the cross sections that govern their interaction rate and final-state kinematics. Through correlations for shared sources of modeling uncertainty, the high-statistics measurement of ν_μ (see Sec. VB) is used to update the ν_e prediction and constrain the total systematic uncertainty. When considering all uncertainties (cross-section uncertainties only), the level of correlation between ν_μ and ν_e events in the signal region is 60% (80%). Throughout this analysis, the ν_μ flux is assumed to be unoscillated.

The ν_μ constraint is implemented by relying on the covariance between ν_μ and ν_e bin contents, the predicted bin content in the different channels, and the observed ν_μ data, and leveraging the properties of block matrices [65]. Given the bin-to-bin covariance matrices for the ν_μ channel ($C^{\mu\mu}$), ν_e channel (C^{ee}), and the covariance between the two channels ($C^{e\mu}$), the predicted bin content in each bin i , m_i^e and m_i^μ for ν_e and ν_μ respectively, and the ν_μ observed bin contents n_i^μ , the constrained ν_e prediction is expressed as:

$$m^e \text{ constrained} = m^e + C^{e\mu} (C^{\mu\mu})^{-1} (n^\mu - m^\mu), \quad (4)$$

and the constrained covariance matrix as:

$$C^{ee} \text{ constrained} = C^{ee} - C^{e\mu} (C^{\mu\mu})^{-1} C^{\mu e}. \quad (5)$$

The fractional systematic uncertainty is presented in Fig. 10 for the three channels included in this analysis (ν_μ , $1eNp0\pi$, and $1e0p0\pi$). The ν_μ selection has no requirement on hadron multiplicity, so the data it selects can constrain both the $1eNp0\pi$ and $1e0p0\pi$ prediction. This selection aims to maximize the reduction of flux uncertainties, particularly at low reconstructed neutrino en-

ergy. Overall, the constraint reduces the systematic uncertainties in the electron neutrino selections by 10–40% relative to the pre-constraint uncertainties. In the analysis, the ν_μ constraint is performed on distributions of the ν_μ and ν_e reconstructed neutrino energy and applied to all quantitative results presented in the next section.

VIII. RESULTS

We first present results from the ν_e selections to test the agreement between the observation and neutrino interaction model prediction and then the tests of the eLEE model. All statistical tests in this analysis are performed over the range 0.15–1.55 GeV in reconstructed neutrino energy with ten 0.14 GeV bins. The test-statistic used is a χ^2 defined as

$$\chi^2 = \sum_{i,j=1}^N (n_i - m_i) C_{ij}^{-1} (n_j - m_j) \quad (6)$$

$$C_{ij} = C_{ij}^{\text{stat CNP}} + C_{ij}^{\text{syst}}, \quad (7)$$

with n_i the observed number of events in bin i , m_i the predicted number of events in bin i for the model being tested, and C_{ij} the covariance matrix, defined in Eq. 7. Statistical uncertainties are the largest in this analysis and are included through the error matrix $C_{ij}^{\text{stat CNP}} = 3/(1/n_i + 2/m_i) \delta_{ij}$ that is constructed using the combined Neyman-Pearson χ^2 definition of Ref. [66]. The systematic error matrix C_{ij}^{syst} is defined in Eq. 2. Toy experiments are generated incorporating systematic variations with a Gaussian sampling of the constrained covariance matrix and then Poisson fluctuating the result to account for statistical variations. For each toy experiment the test-statistic is evaluated, and the distribution is compared to the test-statistic of the data to extract a p -value for the observation. Alternative statistical procedures were also used to validate the results shown in the next sections, and led to similar conclusions as the ones presented.

A. Modeling of electron neutrinos

The observed $1eNp0\pi$ and $1e0p0\pi$ event rates are plotted as a function of reconstructed energy in Fig. 11, where data is compared to the prediction after the ν_μ constraint. Given the agreement observed with the ν_μ selection (Fig. 4), the effect of the constraint procedure on the ν_e prediction is relatively small (less than 10%).

The first statistical test performed is a goodness of fit to quantify how well the intrinsic ν_e model matches the data in reconstructed neutrino energy. The results are presented in Table III. The data are consistent with the intrinsic ν_e model with p -values of 0.182, 0.126, and 0.098 for $1eNp0\pi$ events, $1e0p0\pi$ events, and the two combined, respectively.

TABLE III. Summary of χ^2 and p -value results for the goodness-of-fit tests for the intrinsic ν_e model. The p -values are computed with frequentist studies based on toy experiments.

Channel	χ^2	χ^2/dof	p -value
$1eNp0\pi$	15.2	1.52	0.182
$1e0p0\pi$	16.3	1.63	0.126
$1eNp0\pi + 1e0p0\pi$	31.50	1.58	0.098

Electron neutrino events can be further characterized in terms of their kinematics. Figure 12 shows the angle (θ) of the reconstructed electron candidate relative to the beam direction and the kinetic energy of the leading proton for $1eNp0\pi$ events. Considering only normalization, in the 0.15–1.55 GeV range, 53 events are observed by the $1eNp0\pi$ channel with a constrained prediction of 78.9 ± 11.6 events, corresponding to a deficit of 1.7σ . The deficit is most pronounced at intermediate energies and in the forward direction. As demonstrated by the p -value obtained by the goodness of fit test, the estimated combined statistical and systematic uncertainties accommodate the observed difference.

In the $1e0p0\pi$ channel we observe good overall normalization agreement between data and simulation, with 31 events observed compared to a constrained prediction of 27.8 ± 4.4 , but the simulation under-predicts the data in the energy bins corresponding to $0.15 < E_{\text{reco}} < 0.43$ GeV. In multiple π^0 enriched sideband regions, however, the data is consistent with the prediction within statistical and systematic uncertainties as reported in Sec. VI. In the $0.15 < E_{\text{reco}} < 0.43$ GeV energy bins of the NuMI sideband, the high electron neutrino purity and satisfactory agreement between data and simulation validate the interaction model for low energy electron neutrinos. We find that data events in this range are dominated by single-shower events with a dE/dx profile consistent with a minimum ionizing particle, as expected for both signal and most surviving background events from simulation. Figure 13 shows the angular distribution of $1e0p0\pi$ events, both over the full energy range and in the region corresponding to 0.15–0.43 GeV. Integrated over the full energy range, the angular distribution shows good agreement with simulation. In the low energy bins, where the simulation under-predicts the observed data, events primarily populate the region with $\cos\theta > 0$. More data will be needed to further study these events. The observation of good shape agreement between the data and the prediction in the leading proton kinetic energy distribution shown in Fig. 12b, as well as a visual scan of selected events, suggests that the migration between the $1eNp0\pi$ and $1e0p0\pi$ selections is minimal.

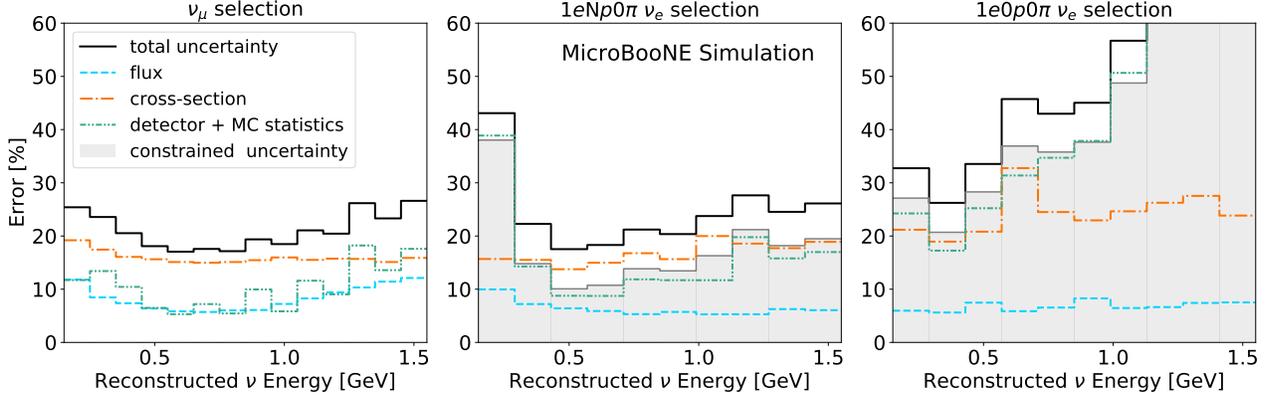


FIG. 10. Summary of the impact of systematic uncertainties presented in Sec. III for all selected events in the three channels used in the analysis, shown in the 0.15–1.55 GeV energy range that is used for quantitative results. The percent systematic uncertainty by channel is shown in the top panel. The individual unconstrained contributions coming from flux, cross section, and detector plus simulation statistics are shown in blue, orange, and green respectively. Detector uncertainties account for both Geant4 re-interaction and detector response modeling uncertainties. The total unconstrained systematic uncertainty is in black and, for the $1eNp0\pi$ and $1e0p0\pi$ selections, a grey area indicates the total constrained uncertainty. The ν_μ uncertainties are not changed by the constraint. In the high energy region of the $1e0p0\pi$ energy spectrum, where we have very few events, the uncertainties grow up to $\mathcal{O}(100\%)$.

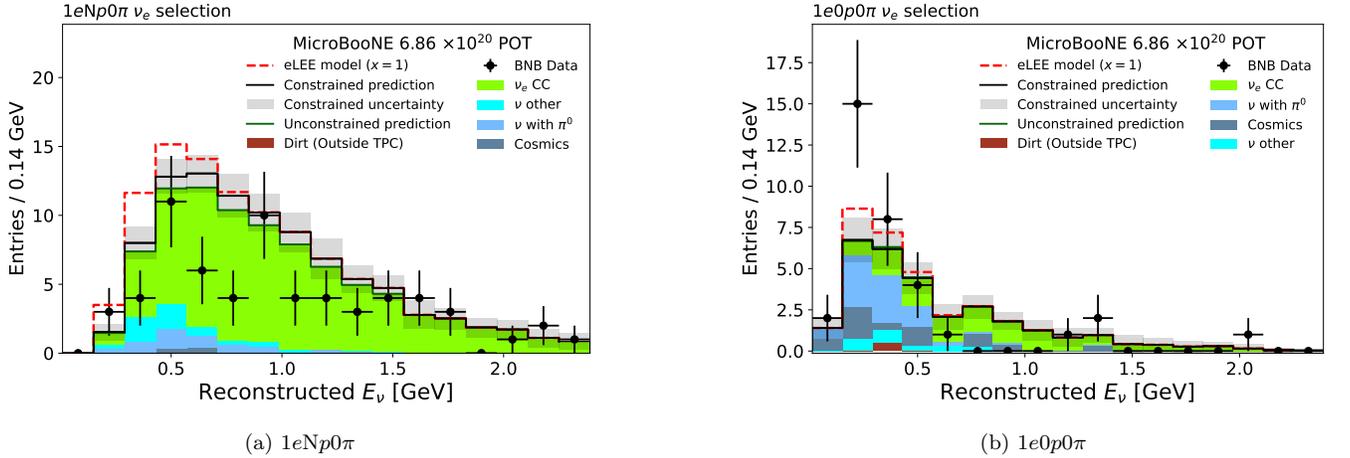


FIG. 11. Reconstructed neutrino energy for the selected $1eNp0\pi$ (a) and $1e0p0\pi$ (b) events. The pre-constraint number of predicted events is shown broken down by true interaction topology. The constrained prediction using the muon neutrino data is also shown both with (red) and without (black) the eLEE model included. Systematic uncertainties on the constrained prediction are shown as a shaded band. While not shown in the figure, systematic uncertainties on the eLEE model are considered in the analysis. Quantitative results are calculated in 10 bins from 0.15 to 1.55 GeV, shown here starting in the second bin. The lower bound of the first bin is 0.01 GeV.

B. Test of the eLEE model

The statistical tests performed to assess the probability that the eLEE model introduced in Sec. III is present in the data are described in this section. The first is a simple hypothesis test in which two hypotheses are tested in order to assess the probability of rejecting one hypothesis assuming the other is true. The hypotheses tested are the intrinsic ν_e prediction (H_0) and the intrinsic ν_e

prediction plus the eLEE model contribution (H_1). The result reported from this test is the p -value based on the $\Delta\chi^2$ between H_0 and H_1 defined as:

$$\Delta\chi^2 = \chi^2(H_0) - \chi^2(H_1). \quad (8)$$

The expected sensitivity and the data results are presented in Fig. 14 and summarized in Table IV. In the $1eNp0\pi$ channel, the data are consistent with H_0 with a p -value of 0.285, which corresponds to 28.5% of the

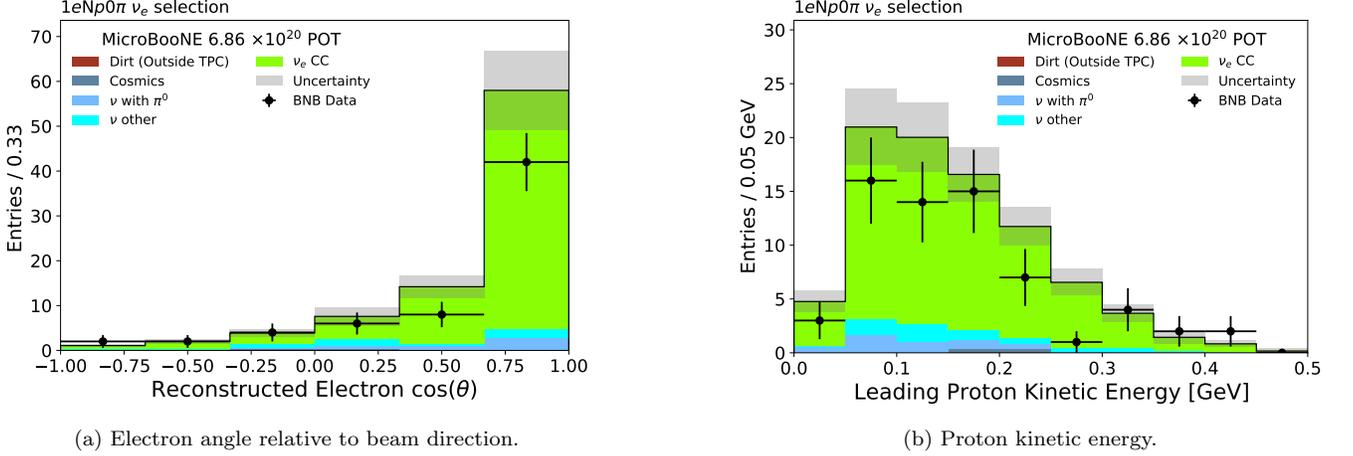


FIG. 12. Selected kinematic distributions for events that pass the $1eNp0\pi$ selection. Expected events and uncertainties are shown as predicted by the nominal simulation.

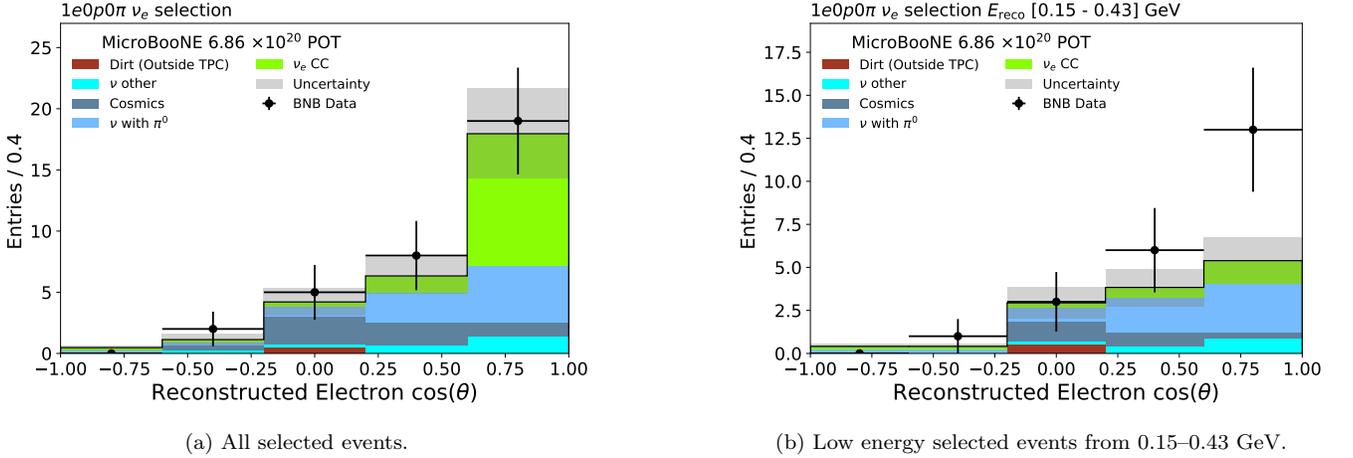


FIG. 13. Selected $1e0p0\pi$ events as a function of electron angle with respect to the beam. Expected events and uncertainties are shown, without the ν_μ constraint applied.

toy experiments that assume H_0 is true showing a $\Delta\chi^2$ smaller than the observed value. When the inverse test is performed, with toy experiments assuming H_1 is true, we find a p -value of 0.021 thus implying that the $1eNp0\pi$ channel alone excludes the H_1 hypothesis at the 97.9% confidence level (CL). In the $1e0p0\pi$ channel, as shown in Fig. 14b, the observed $\Delta\chi^2$ falls in the tail of the expected distribution from both hypotheses. The observation indicates a preference for the H_1 over the H_0 hypothesis with a fraction of toys in the tail of 0.072 for H_1 and of 0.016 for H_0 . While the combined results are expected to be driven by the larger sensitivity of the $1eNp0\pi$ channel to the model tested, the preference for H_1 in the $1e0p0\pi$ channel leads to an intermediate result between the two hypotheses.

The Feldman-Cousins procedure [67] is used to test the signal-strength μ , where μ is a flat scaling parameter

of the eLEE model, and is intended to provide further quantitative insight into a possible signal enhancement at low energy. Toy experiments are generated for different values of true signal-strength μ . In this test the metric for defining the ordering rule based on the likelihood ratio $R(x|\mu)$ is approximated as

$$R(x|\mu) \sim \Delta\chi^2(x|\mu) = \chi^2(x, \mu) - \chi^2(x, \mu_{\text{BF}}) \quad (9)$$

where μ_{BF} is the value of μ that maximizes the likelihood ratio for a given toy-experiment x . Given the observed data, $\chi^2(\text{data}, \mu)$ values are computed for all values of μ and the best-fit value is identified. Confidence intervals are extracted based on the fraction of the toy experiments that give a larger $\Delta\chi^2(x|\mu)$ than $\Delta\chi^2(\text{data}|\mu)$. Results are shown in Fig. 15. Intervals at the 90% CL are reported in Table V where the best-fit value μ_{BF} and the expected sensitivity are also reported. In the $1eNp0\pi$

TABLE IV. Summary of the simple hypothesis tests. Reported p -value(H_0) (p -value(H_1)) results reflect the probability for the H_0 (H_1) hypothesis to give $\Delta\chi^2 = \chi^2(H_0) - \chi^2(H_1)$ smaller than the observed value. The observed value of $\chi^2(H_0)$ is reported in Table III. The median sensitivity in terms of these p -values is also reported under the assumption that the eLEE model H_1 (no-signal scenario H_0) is true. The fraction of toy experiments generated under the H_0 hypothesis with $\Delta\chi^2$ larger than the median value obtained for the eLEE model H_1 is $1 - p$ -value(H_0) so the combined $1eNp0\pi + 1e0p0\pi$ median sensitivity to reject H_0 if H_1 is true is 0.032.

Channel	obs. $\Delta\chi^2$	$\Delta\chi^2 < \text{obs.}$ p -value(H_0)	$\Delta\chi^2 < \text{obs.}$ p -value(H_1)	Sensitivity p -value(H_0) H_1	Sensitivity p -value(H_1) H_0
$1eNp0\pi$	-3.89	0.285	0.021	0.957	0.061
$1e0p0\pi$	3.11	0.984	0.928	0.759	0.249
$1eNp0\pi + 1e0p0\pi$	-0.58	0.748	0.145	0.968	0.049

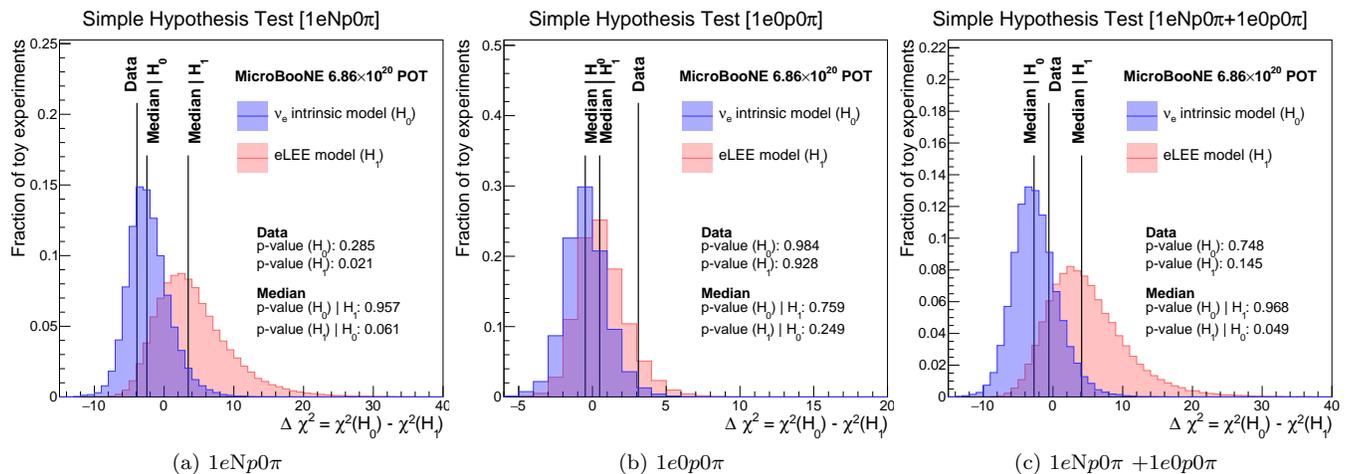


FIG. 14. Results for the simple hypothesis test in the $1eNp0\pi$ (a), $1e0p0\pi$ (b), and combined (c) channels. The $\Delta\chi^2$ between the intrinsic ν_e model and the eLEE hypotheses is plotted for toy experiments generated with these hypotheses. The p -values indicate the fraction of toy experiments with $\Delta\chi^2$ smaller than the observation. The median p -value for toy experiments produced assuming the intrinsic ν_e model is also reported.

channel we find that μ_{BF} is 0, and values of $\mu > 0.82$ are excluded at the 90% CL. Due to the low sensitivity to the eLEE model in the $1e0p0\pi$ channel we find that the 90% confidence interval covers a wide range of μ values, from 1.1 to 15.0. The combined measurement excludes $\mu > 1.57$ at the 90% CL.

Overall, the data are consistent with the intrinsic ν_e model, as shown in Sec. VIII A, but an enhancement of the event topologies measured in the $1e0p0\pi$ channel cannot be ruled out. The data in the separate $1eNp0\pi$ and $1e0p0\pi$ channels suggest that a simple energy-dependent scaling of the ν_e beam content as defined in the eLEE model tested is not favored.

IX. CONCLUSIONS

This article presents a measurement of charged current ν_e interactions without final-state pions in the MicroBooNE detector from the Fermilab BNB. This analysis incorporates numerous sidebands to validate the modeling of the detector response as well as the neutrino flux

and interaction model. Good compatibility between data and simulation is found in all validation data sets, including the ν_μ selection used to constrain the flux and cross section uncertainties of the intrinsic electron neutrino interactions. Electron neutrino interactions are observed with high purity leveraging the power of the MicroBooNE LArTPC detector and are found to be consistent with the ν_e prediction through a goodness of fit test at the 10–20% level. These events are further characterized using their measured kinematic properties of electron angle and kinetic energy of the leading proton.

Comparison to a signal model based on the median MiniBooNE eLEE observation is also studied. When the presence of the eLEE model is tested against the intrinsic electron neutrino interaction model, data in the two channels combined ($1e0p0\pi + 1eNp0\pi$) does not indicate a strong preference between the two hypotheses. The eLEE signal model is disfavored by the $1eNp0\pi$ channel at the 97.9% CL. The eLEE model is further parametrized in terms of the signal strength μ , and confidence intervals for this parameter are extracted with the Feldman-Cousins procedure. The $1e0p0\pi$ selection,

TABLE V. Best-fit eLEE model signal strength (μ) and 90% confidence intervals. The sensitivity is quantified by reporting the expected upper limits assuming $\mu = 0$.

Channel	Data	Data	Sensitivity
	μ_{BF}	90% CL interval on μ	90% upper limit on μ
$1eNp0\pi$	0.00	[0.00 , 0.82]	1.16
$1e0p0\pi$	4.00	[1.13 , 15.01]	3.41
$1eNp0\pi + 1e0p0\pi$	0.36	[0.00 , 1.57]	1.07

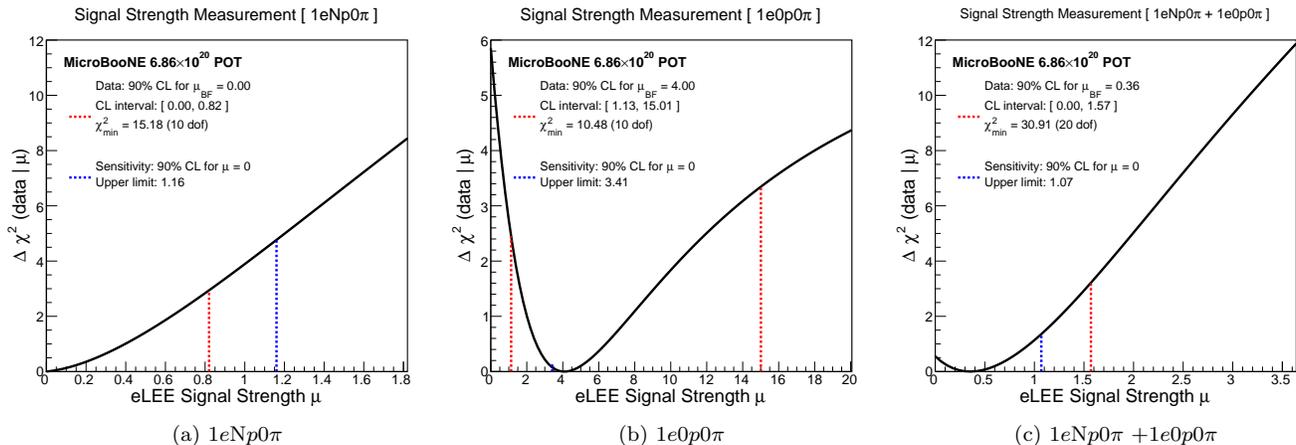


FIG. 15. Results for the signal strength test in the $1eNp0\pi$ (a), $1e0p0\pi$ (b), and combined (c) channels. The $\Delta\chi^2$ as a function of the signal strength is evaluated with respect to the best-fit signal strength value. The observed confidence interval at 90% confidence level is indicated with a vertical lines, as well as the expected upper limit in case of no signal.

which is overall less sensitive to the eLEE model, observes more events than predicted in the lowest energy region and we find that the 90% confidence interval covers a wide range of μ , with a lower bound of 1.1. The $1eNp0\pi$ selection, which drives the analysis sensitivity due to its higher statistics and purity, indicates a preference for no excess of low-energy electron neutrinos with respect to the intrinsic beam content prediction resulting in an upper 90% CL limit on the signal strength of 0.82. More data and tests of additional models will provide further insight into these results.

This analysis is part of a broad effort by the MicroBooNE collaboration to measure low-energy electromagnetic interactions. It will be followed by additional analyses, including those which use the full MicroBooNE dataset, roughly twice the size of that used in this result as well as advances in low energy shower reconstruction and analysis tools. Finally, the full SBN program [68], with the introduction of a near detector and a third detector at a longer baseline, will further expand this investigation.

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