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Ionization electron signal processing in single phase LArTPCs. Part II. Data/simulation comparison and performance in MicroBooNE

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Abstract: The single-phase liquid argon time projection chamber (LArTPC) provides a large amount of detailed information in the form of fine-grained drifted ionization charge from particle traces. To fully utilize this information, the deposited charge must be accurately extracted from the raw digitized waveforms via a robust signal processing chain. Enabled by the ultra-low noise levels associated with cryogenic electronics in the MicroBooNE detector, the precise extraction of ionization charge from the induction wire planes in a single-phase LArTPC is qualitatively demonstrated on MicroBooNE data with event display images, and quantitatively demonstrated via waveform-level and track-level metrics. Improved performance of induction plane calorimetry is demonstrated through the agreement of extracted ionization charge measurements across different wire planes for various event topologies. In addition to the comprehensive waveform-level comparison of data and simulation, a calibration of the cryogenic electronics response is presented and solutions to various
MicroBooNE-specific TPC issues are discussed. This work presents an important improvement in LArTPC signal processing, the foundation of reconstruction and therefore physics analyses in MicroBooNE.

**KEYWORDS:** Neutrino detectors; Performance of High Energy Physics Detectors; Data processing methods; Time projection Chambers (TPC)

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1 Introduction

MicroBooNE [1], or the Micro Booster Neutrino Experiment, is located at Fermilab in Batavia, Illinois, and has the goal of addressing the low-energy electron-neutrino-like excess observed at MiniBooNE [2] on the same beam line (the Booster Neutrino Beam, or BNB). In order to address this anomaly, the experiment utilizes a different detector technology, a liquid argon time projection chamber (LArTPC). The LArTPC [3–6] provides fine-grained spatial resolution and an excellent calorimetric capability that allows for discrimination between electrons and photons in the active volume of the TPC [7]. This discrimination power is key to the success of the MicroBooNE physics program, as it allows one to identify the photon backgrounds which complicate the MiniBooNE electron appearance search.

The MicroBooNE LArTPC, illustrated in figure 1, utilizes three wire planes at the anode to obtain signals from drifting ionization electrons produced from neutrino interactions within
the active volume of the detector. Drifting ionization electrons first pass by the two induction planes (the “U plane” and “V plane”), inducing signals on the wires, and are finally collected on the collection plane (the “Y plane”). The drift electric field is approximately 273 V/cm and is maintained by a cathode held at −70 kV as well as a field cage that surrounds the LArTPC active volume and gradually steps down the voltage toward the anode plane via a resistor chain. The three wire planes are biased at specific voltages (−110 V, 0 V, and +230 V for the U plane, V plane, and Y plane, respectively) to ensure complete transparency for drifting ionization electrons, allowing for all ionization electrons to reach the third (collection) wire plane without being collected by one of the first two (induction) wire planes. Bipolar signals are produced on the induction planes and largely unipolar signals are produced on the collection plane.

One primary asset of the LArTPC technology is its capability of discriminating between electrons and photons in the detector. This is strongly desired in accelerator neutrino experiments as photons are often backgrounds to searches for electron neutrino appearance. Photons that convert into an electron-positron pair within the liquid argon can be identified as having roughly double the value of $dE/dx$ compared to a single electron, measured using the calorimetric information (recovered from the charge measured at the TPC wires) of the track-like beginning of the particle signature in the detector. In order to make use of the electron versus photon discrimination capability of the LArTPC technology, it is necessary to first demonstrate the ability to extract charge precisely from these signals on the TPC wires; both imaging and calorimetry in the TPC active volume depend on this precise charge extraction.

Three elements come into play for signal formation on TPC wires: i) the distribution of the ionization charge cloud produced by charged particles traveling through the TPC; ii) the field response describing charge collection and induction on the TPC wires as ionization electrons drift through the wire planes; and iii) the electronics response characterizing the amplification and shaping of the induced current on the TPC wire in the front-end electronics. A detailed simulation of each of these effects is required to realistically simulate particle interactions within the MicroBooNE detector for comparison with data. GEANT4 [9] is used to simulate the energy deposition of charged particles traveling through the liquid argon and LArSoft [10] is utilized to simulate ionization charge recombination, transport, and diffusion. For the wire field response and electronics amplification and shaping, a dedicated simulation was developed for MicroBooNE [11]. The electronics response shape is taken from a parametrization of the expected response of the front-end electronics at a given gain and peaking time, while the wire field response is simulated using Garfield [12]. This wire field response simulation includes a dependence on the impact parameter of the ionization electron drift with respect to the wires, as well as the charge induction on wires both close to and far from the drifting ionization electrons.

After the employment of software noise filtering techniques to eliminate excess noise from external sources [13], a robust signal processing chain [11, 14] is required to extract the number of drifting ionization electrons passing through each wire plane. It is necessary to validate this signal processing chain with data from the MicroBooNE LArTPC in order to evaluate the performance of drifted charge extraction from the TPC wires. This work describes this validation, as well as the validation of the TPC signal simulation described in ref. [11], by making comparisons between MicroBooNE data and simulation. In addition, the performance of charge-matching across the three wire planes of the MicroBooNE LArTPC is illustrated in this work. This charge-matching can be
utilized in novel imaging techniques [15] that take advantage of local cross-plane charge correlations in order to construct three-dimensional images. This approach simplifies charge clustering for electron and photon showers and is expected to help in the reconstruction of neutrino interactions.

This article is organized as follows. First, in the remainder of the introduction, a brief review of signal processing concepts referenced throughout this work is presented. In section 2 the calibration of the electronics response function on a channel-by-channel basis, which enables a uniform electronics response across all TPC channels, is described. Next, in section 3, a comparison of data and simulation at the waveform level is carried out, characterizing the capability of the ionization signal simulation to reproduce the response of the TPC wires and the electronics to drifted ionization charge. This section also includes discussion of a method used to account for the modified response function in the shorted regions of the detector, where a group of channels in the collection plane (Y plane) are shorted to a non-standard bias voltage. Special considerations necessary to employ the signal processing chain discussed in ref. [11] on data are presented in section 4, including the addressing of electronics/noise issues that complement the studies highlighted in ref. [13]. Results of applying the signal processing chain on MicroBooNE data are shown in section 5, including the characterization of its performance on events in data and a demonstration of the precise extraction of charge information from the induction planes of a

Figure 1. Illustration of signal formation in the MicroBooNE three-plane LArTPC, depicting the V plane (second induction plane) and Y plane (collection plane) TPC wire signals on the right of the image. Reproduced from [8]. © 2017 IOP Publishing Ltd and Sissa Medialab srl.
LArTPC, which has previously been shown to be feasible [16]. Emphasized in this section is the matching of charge across the three planes of the MicroBooNE LArTPC. Finally, a summary of the investigations carried out in this work, along with their significance, is presented in section 6.

1.1 Review of signal processing concepts

A number of fundamental signal processing concepts are referenced throughout this article that are worth discussing briefly as background information. For a more complete discussion of LArTPC signal processing concepts and methodology, see ref. [11].

The deconvolution technique is used to extract charge information from LArTPC waveform data, and has been utilized previously in the data analysis of other LArTPC experiments [14, 17]. More advanced deconvolution techniques have been explored in MicroBooNE [11] and are validated with data events in this work. In brief, the deconvolution is used to extract the original ionization charge signal \( S(t) \) from the measured signal \( M(t') \). The measured signal is modeled as the convolution of the original signal \( S(t) \) and a known detector response function \( R(t' - t) \):

\[
M(t') = \int_{-\infty}^{\infty} R(t' - t) \cdot S(t) \, dt.
\]

This “full” detector response \( R(t' - t) \) includes both the wire field response and the electronics response, and can be either simulated or estimated from data using an independent calibration sample (see section 2 and section 3). One can transform eq. (1.1) into the frequency domain by applying a Fourier transformation, yielding \( M(\omega) = R(\omega) \cdot S(\omega) \), where \( \omega \) is in units of angular frequency. The original signal in the frequency domain \( S(\omega) \) can be found by rearranging this equation:

\[
S(\omega) = \frac{M(\omega)}{R(\omega)}.
\]

This solution is extended to account for additional detector and electronics noise contributions through the introduction of a low-pass filter function \( F(\omega) \), yielding

\[
S(\omega) = \frac{M(\omega)}{R(\omega)} \cdot F(\omega).
\]

The purpose of the filter function is to attenuate problematic high-frequency noise that is amplified by the deconvolution procedure (division of noise by the response function at frequencies where the signal response is very small).

The above deconvolution prescription is known as the “one-dimensional (1D) deconvolution” in the sense that a fast Fourier transform (FFT) is carried out in the time dimension only. This procedure assumes the induced current on each sense wire is independent of the topology of the charge distribution creating the signals on the wires. However, as described in ref. [11], the induced current on the sense wire receives additional topology-dependent contributions from ionization charge drifting past adjacent wires at roughly the same time. Accounting for these additional contributions and assuming the field response strength between wires is approximately constant allows eq. (1.1) to be generalized as

\[
M_i(t_0) = \int_{-\infty}^{\infty} \left( \ldots + R_1(t_0 - t) \cdot S_{i-1}(t) + R_0(t_0 - t) \cdot S_i(t) + R_1(t_0 - t) \cdot S_{i+1}(t) + \ldots \right) \, dt,
\]
where \( M_i \) represents the measured signal from wire \( i \). \( S_{i-1}, S_i, \) and \( S_{i+1} \) represent the real signal inside the boundaries of wire \( i \) and its adjacent intra-plane neighbors. \( R_0 \) and \( R_1 \) are the average full response functions for ionization charge passing through the wire region of interest and adjacent wire regions respectively.

Converting this equation into the frequency domain by applying a Fourier transform yields

\[
M_i(\omega) = \ldots + R_1(\omega) \cdot S_{i-1}(\omega) + R_0(\omega) \cdot S_i(\omega) + R_1(\omega) \cdot S_{i+1}(\omega) + \ldots, \tag{1.5}
\]

which can be written in matrix notation as

\[
\begin{pmatrix}
M_1(\omega) \\
M_2(\omega) \\
\vdots \\
M_{n-1}(\omega) \\
M_n(\omega)
\end{pmatrix} =
\begin{pmatrix}
R_0(\omega) & R_1(\omega) & \ldots & R_{n-2}(\omega) & R_{n-1}(\omega) \\
R_1(\omega) & R_0(\omega) & \ldots & R_{n-3}(\omega) & R_{n-2}(\omega) \\
\vdots & \vdots & \ddots & \vdots & \vdots \\
R_{n-2}(\omega) & R_{n-3}(\omega) & \ldots & R_0(\omega) & R_1(\omega) \\
R_{n-1}(\omega) & R_{n-2}(\omega) & \ldots & R_1(\omega) & R_0(\omega)
\end{pmatrix}
\begin{pmatrix}
S_1(\omega) \\
S_2(\omega) \\
\vdots \\
S_{n-1}(\omega) \\
S_n(\omega)
\end{pmatrix}. \tag{1.6}
\]

Inverting the matrix \( R \) via application of Fourier transforms allows for the determination of the original signal vector \( S \). This “two-dimensional (2D) deconvolution” yields a more accurate recovery of the ionization electron distribution using the known detector response in both the time and wire dimensions.

The 2D deconvolution procedure described above provides a robust and computationally efficient method to extract the distribution of ionization electrons reaching each wire at the LArTPC anode. While this procedure works well for the collection plane, additional signal processing is necessary for the induction planes due to the suppression of the induction plane wire response at low frequencies, a feature associated with the bipolar signals observed on the waveforms. This suppression leads to low-frequency noise becoming substantially amplified in the 2D deconvolution. Without mitigation, the amplification of low-frequency noise would lead to large uncertainties in the estimation of charge induced on the waveforms. In order to address this issue, signal regions of interest (ROIs) are first selected on the induction plane waveforms using high-pass filters that reduce the impact of the low-frequency noise when finding the signals. Once the ROIs are found, the local baseline of the signal contained in the ROI is defined by an interpolation using the signal sidebands, and a Gaussian filter is utilized to extract the ionization charge. This procedure is discussed in more detail in ref. [11].

2 Calibration of TPC cold electronics response function

Each wire is individually instrumented by an electronic circuit that amplifies and shapes its current signals [18]. These pre-amplifiers are implemented as application-specific integrated circuits (ASICs) and are operated at cryogenic temperature inside the liquid argon cryostat in order to reduce electronic noise and simplify detector cabling. The amplified analog voltage signals are routed out of the detector cryostat, amplified again by intermediate warm electronics and digitized at 2 MHz by 12-bit analog to digital converters (ADCs) [8].

Cryogenic (cold) pre-amplifying electronics are a crucial element of the MicroBooNE LArTPC as the reduced electronic noise during cryogenic operation compensates for the absence of ionization
charge amplification within the liquid argon. Charge measurements made with induction wire signals especially require low electronic noise due to the relatively small currents induced on the wires by the drifting ionization charge [19]. Excellent cold electronic noise performance was observed in the MicroBooNE detector following the start of regular operation, with an equivalent noise charge (ENC) measured to be less than or equal to $400e^-$ for the majority of channels compared to an expected signal size of $\sim 17000e^-$ from minimum ionizing particles traveling parallel to the wire plane and perpendicular to the wire direction [13].

Each individual cold electronics ASIC channel can be configured to use one of four different gain settings ($4.7, 7.8, 14$ and $25$ mV/fC) and peaking time settings ($0.5, 1, 2$ and $3$ μs). During normal operation, MicroBooNE uses the 14 mV/fC gain setting and 2 μs peaking time, which satisfies the Nyquist criterion [20] for 2 MHz digital sampling. The output baseline voltage is set to $\sim 200$ mV and $\sim 900$ mV for the collection plane and induction planes, respectively, in order to maximize the dynamic range of signals on the waveform. The baseline voltages are well-matched to both the overall 1.6 V dynamic range of the cold electronics and the 2 V range of the ADCs.

This section describes the in-situ calibration of the MicroBooNE LArTPC cold electronics using an externally generated signal injected into the pre-amplifier channel inputs. Calibrating the cold electronics response allows for a direct comparison between the ideal and real electronics responses. This in-situ calibration can also be used to improve the accuracy of simulated data and signal processing algorithms. It is also of interest to determine the uniformity of the electronics response. Finally, repeating the calibration measurement periodically is required in order to evaluate the stability of the electronics response over time.

2.1 MicroBooNE TPC cold electronics calibration system

The MicroBooNE cold electronics calibration system uses a signal generator located outside the cryostat to provide a calibration signal to each individual cold electronics channel input, as summarized in figure 2. The signal generator produces a square wave signal that is fanned out and propagated into the detector cryostat through dedicated service cables. Service cables inside the cryostat connect to the ASIC carrier motherboards where the calibration signal is routed to each individual ASIC test input pin. Internally, the ASIC capacitively couples the test input signal to each pre-amplifier input individually through a $\sim 183$ fF capacitor. The ASICs must be configured to connect the test signal input pin to the calibration capacitors; during normal detector operation they are disconnected. The capacitively coupled square wave calibration signal injects a known amount of charge into the pre-amplifier input on a timescale of $\sim 10$ ns, which is much shorter than the nominal pre-amplifier peaking time of 2 μs. Consequently the resulting signal waveform approximates the impulse response of the cold pre-amplifiers. An example of a digitized calibration signal waveform is shown in figure 3.

A number of channels of the cold electronics are considered non-functional and cannot be calibrated, with the specific causes discussed in ref. [13]. Roughly 220 channels (out of 8256) of the cold electronics are located on ASICs that cannot be configured due to a hardware failure of the configuration chain from a service board through the cold cable to the cold electronics inside the cryostat. These channels are still used to measure ionization charge but can only use the default 4.7 mV/fC gain and 1 μs peaking time settings. Calibration signals can not be injected into these channels as the ASICs must be specifically configured to enable the test input pin. An additional
**Figure 2.** Schematic overview of the MicroBooNE cold electronics calibration signal system. The plot shows a $\sim 50$ mV square wave voltage signal, generated outside the LAr cryostat and propagated to the cold electronics channel inputs via the calibration injection system, shown schematically in the graphic. Copies of the square wave voltage signal are generated with a fanout module and transmitted into the cryostat through dedicated service boards and cables. The signal is routed to every cold electronics ASIC test input pin and coupled to each pre-amplifier input through a $\sim 183 \, \text{fF}$ capacitor.

**Figure 3.** Example of a digitized waveform from an injected calibration signal on a single collection plane channel, including a fit to the electronics response function (overlaid in red). The fitted peaking time $t_p$ is 2.15 $\pm$ 0.01 $\mu$s and the amplitude $A_0$ is 599 $\pm$ 3 ADCs.
∼ 860 channels are not functioning correctly and similarly can not be calibrated. These channels are either unresponsive due to ASIC failure or to a short between different wire planes, and are approximately stable in number over time as discussed in ref. [13].

2.2 Electronics response parameterization

The impulse response of the TPC cold electronics is well described by a fifth-order semi-Gaussian anti-aliasing filter. The representation of this function in the time domain is shown in figure 3 and is parameterized by

\[ R(t, A_0, t_p) = A_1 E_1 - A_2 E_2 (\cos \lambda_1 + \cos \lambda_1 \cos \lambda_2 + \sin \lambda_1 \sin \lambda_2) \]
\[ + A_3 E_3 (\cos \lambda_3 + \cos \lambda_3 \cos \lambda_4 + \sin \lambda_3 \sin \lambda_4) \]
\[ + A_4 E_2 (\sin \lambda_1 - \cos \lambda_2 \sin \lambda_1 + \cos \lambda_1 \sin \lambda_2) \]
\[ - A_5 E_3 (\sin \lambda_3 - \cos \lambda_4 \sin \lambda_3 + \cos \lambda_3 \sin \lambda_4). \]

(2.1)

The parameters in eq. (2.1) are obtained from a detailed simulation of the filter design and given by:

\[ A_1 = 4.31054 A_0, \quad A_2 = 2.6202 A_0, \]
\[ A_3 = 0.464924 A_0, \quad A_4 = 0.762456 A_0, \quad A_5 = 0.327684 A_0, \]
\[ E_1 = e^{-2.84999 t_p / t_p}, \quad E_2 = e^{-2.45533 t_p / t_p}, \quad E_3 = e^{-2.40318 t_p / t_p}, \]
\[ \lambda_1 = 1.19361 t_p, \quad \lambda_2 = 2.38722 t_p, \]
\[ \lambda_3 = 2.5928 t_p, \quad \lambda_4 = 5.18561 t_p, \]

(2.2)

where \( t \) is time in μs, \( t_p \) is the peaking time constant in μs, and \( A_0 \) is the amplitude parameter in units of ADC counts for digitized waveforms.

Electronics response calibration data are recorded while the detector is in its normal operational state. The cold electronic pre-amplifier ASICs are configured to use their default settings with the exception that the calibration signal input is enabled for every electronics channel in the detector. Calibration signal pulse shapes are identified in digitized waveform data and the leading edge fitted using the response function described in eq. (2.1). Calibration signal amplitude and peaking time parameters for each channel are obtained from these fits. The results of these fits are used to parameterize the electronics response for each channel.

Electronics channel gain can be derived from the fitted calibration pulse amplitude given the known amplitude of the injected signals. These gain measurements need to be corrected to account for attenuation introduced by the calibration signal injection system. The most important correction to electronics response gain measurements is due to the calibration signal fanout modules. A pair of modules receives the analog calibration signal from the function generator and propagates copies to each set of motherboards sharing common service cables. The calibration signal input for one of these modules is output from the other, and this daisy-chained configuration attenuates the second calibration signal copy. This can be observed directly by comparing calibration signal average pulse height measurements for sets of channels connected to each module, as in the top row of figure 4. A correction is applied to the gain measurements for channels receiving the calibration
module attenuated signal to remove this effect. A similar correction removes variations in gain measurements due to service cable attenuation. These corrections are constant over time, although there are two sets of service cable attenuation corrections corresponding to the periods before and after a service board hardware upgrade was performed in summer 2016. Calibration signal injection is observed to induce electronic crosstalk into adjacent channels with a bipolar signal size \( \sim 1\% \) of the calibration pulse height, which is small enough to not require any correction.

The gain measurements after applying the previously defined calibration module fanout and service cable corrections are shown in the bottom row of figure 4. There is a systematic discrepancy in measured gain between channels configured with the collection baseline setting and those configured with the induction baseline setting, as shown in figure 5, that is an intrinsic feature of the cold pre-amplifiers. Channels configured to use the induction baseline setting have an average measured gain of \( 194.3 \pm 2.8 \) \([e^-/ADC]\) while the average collection-mode channel gain is \( 187.6 \pm 1.7 \) \([e^-/ADC]\). This discrepancy contributes to some of the residual variations in the induction wire corrected gain measurements as a subset of induction plane channels are misconfigured to use the
collection 200 mV baseline setting. The variation in the ~ 183 fF test signal injection capacitors is ~ 0.5%, which contributes to the observed variation in these gain measurements. Additionally, these gain measurements are also expected to have a systematic bias due to the overall average attenuation introduced by the calibration injection system. Calibration methods such as measuring energy deposition from minimum ionizing particles are required to measure the absolute gain given the inability to measure this overall attenuation factor in-situ.

The electronics response peaking time parameter is obtained from fits to calibration signals for each cold pre-amplifier channel individually. Peaking time parameters are not measured for misconfigured or non-functioning channels. The mean peaking time is 2.18 ± 0.08 μs, which is ~ 10% higher than the nominal value of 2 μs, due to the additional input capacitance from the TPC wires. The overall variation in fitted pre-amplifier peaking time is ~ 3%, which is significantly less uniform than the electronics gain with ~ 1% variation. This variation does not seem to be caused by the implementation of the calibration system but is a feature of the pre-amplifier ASICs. Figure 6 suggests that the source of this variation is a disagreement in the shape of the ideal and real electronics response. In particular, the real electronics response has much longer tails than does the ideal response, which is a known feature in this version of the cold electronic ASICs and has been addressed in a subsequent version of the electronics [21].

The observation of non-ideal components in the calibration signal waveforms suggests that parameterizing the response in terms of the idealized response may introduce bias in charge measurements when comparing data to simulation. As an alternative, it is possible to measure the average impulse electronics response shape from calibration signals and use this in place of the ideal response for use in signal processing algorithms, as described in section 2.4.
Figure 6. Example calibration signal waveforms, averaged over many signal pulses, with clearly visible extended tails. These extended tails are not well described by the ideal electronics response, which is also shown for comparison (black). This is a known feature in this version of the cold electronic ASICs and has been addressed in a subsequent version of the electronics [21]. These tails are corrected in the signal processing chain using the deconvolution-based method described in this section.

2.3 Stability of electronics response

In order to apply electronics response measurements in signal processing algorithms, it is necessary to demonstrate that the response is stable over time. If the response changes significantly over time, there is the potential to bias ionization charge measurements by using response parameters that are not appropriate for a given time period. The stability of the electronics response is evaluated by comparing the measured electronics response parameters from calibration runs spaced several months apart. It is observed that the measured cold pre-amplifier gain is very stable over this timescale, as shown in figure 7, with an average deviation less than 0.2% across all measurements.

2.4 Validation of electronics response correction

In this section we demonstrate how the measured average response waveforms can correct the observed non-uniform and non-ideal electronics response. This correction is achieved through a 1D deconvolution based on discrete-space Fourier transformation techniques described in section 1.1. The correction can be defined in the frequency domain as:

$$M^\text{corr}_i(\omega) = M_i(\omega) \cdot \frac{R_{\text{ideal}}(\omega)}{R_i(\omega)},$$

(2.3)

where $\omega$ is in units of angular frequency, $R_{\text{ideal}}(\omega)$ is the Fourier transform of the ideal electronics response function and $R_i(\omega)$ is derived from the measured average electronics response for the $i$th channel. Since the $R_{\text{ideal}}$ and $R_i$ are generally similar, the change in the electronics noise is minimal and can be safely ignored. The above procedure only acts on a single channel at a
Figure 7. Gain measurements of the TPC cold electronics channels taken over time, with red and blue colored bands representing the RMS variation of the corrected channel gain in the induction and collection planes, respectively. The average discrepancy in channel gain measurements between calibration data runs is less than 0.2% and is significantly smaller than the variation in channel gain within the induction and collection wire planes.

The electronics response correction defined in eq. (2.3) is applied to injected calibration signals in order to evaluate the impact on the shape of the impulse response. Injected calibration signals with and without the electronics response correction applied are fitted with the ideal electronics response function. The distributions of the measured peaking times are summarized in figure 8. The calibration signal shapes are significantly more uniform after the electronics response correction is applied, with ∼0.8% variation in the measured peaking time parameter after the correction compared to the original ∼3% variation.

This correction is also evaluated by applying it to signals produced by a cosmic muon track recorded during normal detector operations as shown in figure 9. This and following event displays show digitized waveform samples from a subset of wires in a specified plane over a fraction of the total time period recorded for the event. The channels and time window were chosen to highlight signals produced by the ionization charge of the cosmic muon, which produce a recognizable image of the track. Signals before and after the response correction are both shown. The horizontal axes correspond to the number of wires relative to the leftmost wire while the vertical axes correspond to units of time. The vertical axes time bins are 3 μs intervals and the contents are the sum of the six digitized samples within that interval with average baseline subtracted, which defines the vertical scale. The relative spatial displacement away from the wire planes of ionization charge producing the observed signals can be inferred from the difference in time and assuming the nominal ionization electron drift velocity. Of particular note in these event displays is that the long tails observed on several collection channels following large ionization charge deposition are removed.
3 Validation of field response function simulation with data

As discussed in section 1, the characteristics of TPC signals are dictated by the initial ionization charge cloud formation within the detector bulk, the signals created on the TPC wires due to the drift of the ionization electrons near the wires (field response), and the shaping of the signals via the front-end electronics (electronics response). In the previous section, the calibration of the electronics response was discussed. This section now focuses on the nature of the wire field response, which must be probed with ionization signals as opposed to signals from the electronics calibration system described in section 2.

In order to validate the simulation of the wire field response and electronics response [11], we compare the simulated field response and electronics response to that seen in data. This requires utilizing a data-driven technique to produce comparisons of simulation and data at the waveform level. In section 3.1, the data-driven method used to make this comparison is described. In particular, comparisons of the full response, which is the convolution of the wire field response and electronics response, are made. Despite making comparisons of the full response, this method primarily demonstrates how well we simulate the wire field response, as the electronics response is first calibrated in data using the cold electronics calibration system (see section 2). The consistency of the wire field response between data and simulation is also an indirect validation of the signal processing chain that is used to “remove” the field and electronics response from the waveform, which is done in order to extract the amount of charge seen on the TPC wires. Correctly modeling the wire field response and electronics response allows for an unbiased estimation of ionization charge on the TPC wires in data.

There is an additional complication due to there being multiple regions within the TPC exhibiting non-standard wire field responses. This is due to the shorting of channels across wire
Figure 9. Example event displays showing the Y plane view from Event 0, Run 3455 (a) before and after the application of the electronic response correction defined in eq. (2.3) in units of average baseline subtracted ADC scaled by 250 per 3 $\mu$s. The long tails (faint vertical stripes going upward from the track) seen on certain collection plane channels following large charge signals are largely removed by the correction. The effect of the correction on an individual digitized waveform (b) also clearly shows this tail removal.

planes in particular regions of the TPC, as described in ref. [13]. The nature of the shorting of channels between the different wire planes has not been observed to change since the beginning of data-taking at MicroBooNE. The different bias voltages seen on channels in these regions lead to different induced current distributions as ionization electrons move past the wires. In addition to the “normal” region, where there are no shorted channels, there is a “shorted-U” region where multiple U plane channels are shorted to one or more V plane channels, and a “shorted-Y” region where multiple Y plane channels are shorted to one or more V plane channels. The wire field responses are extracted separately in these different regions and the response is assumed to be uniform within each of these regions. These subregions are illustrated in figure 10. Similar to the normal region, the field response simulation in the shorted-channel regions is done using Garfield with identical wire geometry as described in ref. [11], in which the signal formation mechanism has been elaborated in detail. In contrast to the simulation in the normal region, in the shorted-channel regions the bias
voltages on the shorted channels are adjusted to yield significant voltage reduction on the wires. This tuning leads to better agreement of the simulated wire field response to that observed in data.

After discussing the data-driven method used to extract the full response from data in section 3.1, comparisons of the full responses in data and simulation are made for the normal region in section 3.2 and for the shorted-U and shorted-Y regions in section 3.3. The detailed Garfield simulation steps are also discussed in section 3.3, focusing on the differences between the simulation in the normal and shorted-channel regions. Finally, in section 3.4 the special treatment of the shorted-Y region in the signal processing chain is discussed.

### 3.1 Response extraction methodology

In order to compare the full response in data to that predicted by simulation, a track-based data-driven method is used to extract the full response from off-beam cosmic ray data (data recorded when it is known that no beam-related neutrinos are passing through the MicroBooNE detector). This is done by i) determining the value of \( t_0 \), where \( t_0 \) refers to the time at which the cosmic ray enters the TPC volume, ii) ensuring that a substantial amount of light is seen by the MicroBooNE photomultiplier tubes (PMTs) [22] at the time corresponding to the value of \( t_0 \) found, iii) correcting the drift coordinate, \( x \), of the cosmic ray track in the TPC to the location the cosmic ray actually traversed through the TPC, and iv) coherently adding the waveforms associated with the \( t_0 \)-tagged track to recover a clean representation of the signal shape from the waveform while suppressing the noise on the waveform. The MicroBooNE PMT system (32 PMTs located behind the anode plane) is used both for this \( t_0 \)-tagging of cosmic tracks as well as for triggering on light associated with beam events within the detector, which is used to enhance the purity of neutrino interaction signal events in the recorded data stream. The motivation for using \( t_0 \)-tagged tracks for the data/simulation comparison comes from the desire to minimize the effects of diffusion in the full response estimation. This can be done by utilizing only parts of tracks very close to the anode plane where effects of diffusion are minimal. One must know the drift coordinate of the tracks in order to ensure this condition, necessitating the use of \( t_0 \)-tagged tracks.

The \( t_0 \)-tagging method used in these studies is visualized in figure 11. First, a sample of data events (on order of 10k) are collected from the MicroBooNE detector using off-beam triggers (data
collected at periodic intervals while the beam is not active), providing a large set of cosmic rays (more than 100k) for study without contamination from neutrino interaction events. The track associated with each cosmic ray is then reconstructed in three dimensions (3D) using the Pandora multi-algorithm pattern recognition [23] employed on TPC data. The cosmic ray is assumed to be through-going (passing through two faces of the TPC). This is a good assumption as most cosmic rays have high enough momentum to pass through the TPC completely without stopping. Next, the track is required to either enter or exit a TPC face in \( y \) (top or bottom of TPC) or \( z \) (front or back of TPC, with respect to direction of the beam), but not both. If this condition is met, and the cosmic ray is through-going (as assumed), the cosmic ray must either enter or exit the anode or cathode. Depending on the angle that the reconstructed track makes in the \( y-x \) plane, the cosmic ray is then determined to be either “anode-piercing” or “cathode-piercing” (see figure 11), and the value of \( t_0 \) is assigned based upon the time tick associated with the signal on the waveform at the part of the track closest to either the anode or cathode, respectively. A “flash” of light seen in several of MicroBooNE’s 32 eight-inch PMTs is required to be found at the same time as the determined \( t_0 \) value in order to increase the purity of the \( t_0 \)-tagging technique. Finally, the drift coordinate of the reconstructed track is adjusted accordingly using the known drift velocity \( (v_{\text{drift}}) \) of ionization electrons in liquid argon at an electric field of 273 V/cm, roughly 1.1 mm/μs [24]. Validations using Monte Carlo simulation and data-driven validations utilizing an external small cosmic ray tagger [25] have shown this \( t_0 \)-tagging method to reconstruct the track \( t_0 \) correctly at least 98% of the time for anode-piercing tracks and at least 97% of the time for cathode-piercing tracks.

With the drift coordinate of the cosmic ray track corrected, one can proceed to use the subset of anode-piercing tracks (roughly one per recorded off-beam event) to estimate the full response

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**Figure 11.** Illustration of the \( t_0 \)-tagging method used in the studies presented in this work. The drift coordinate \( (x) \) of cosmic ray tracks are corrected using the assumption that the cosmic ray is through-going, requiring the track to pass through only one TPC face in \( y \) or \( z \), and using the angle of the track in the \( y-x \) plane (or \( z-x \) plane) to determine if the track is anode-piercing (a) or cathode-piercing (b). The known ionization electron drift velocity, \( v_{\text{drift}} \), then allows for the correction of the drift coordinate of the cosmic ray track. The track before and after the correction is shown by the solid and dashed lines, respectively. Before the correction (when the \( t_0 \) of the track is assumed to be the trigger time), the track falsely appears to stop in the middle of the TPC.
via the method described above. Only waveforms associated with the portion of the track between 2 cm and 10 cm away from the anode plane are used in order to minimize the influence of diffusion, which is why anode-piercing tracks are used instead of cathode-piercing tracks. Waveform signals are lined up in time across many different wires (and tracks from different events) and are added together. This can be done as cosmic muons are well-approximated by straight lines, so the signal can be assumed to be coherent across a short distance of the cosmic muon track. For unipolar signals, such as those observed by the collection plane wires, the positive signal peak is the feature lined up in time; in contrast, for the bipolar signals of the induction plane wires, the negative dip of the signal is instead lined up in time. The maximum (minimum) ADC value in the region of the waveform containing the track is used to line up the collection (induction) signals in time. The coherent nature of the signal across waveforms associated with different wires leads to the signal response being preserved when the waveforms are added together. Conversely, noise on the waveform averages out as it is incoherent across different wires. The result is a data-driven estimate of the full response for each plane in each of the regions.

The coordinate system used for these studies is described in figure 12a. For each wire plane, the $x$-axis is along the drifting field direction (opposite to the electron drift direction), the $y$-axis is along the wire orientation, and the $z$-axis is along the wire pitch direction. Different coordinates are used for each wire plane, as the wire directions of the different planes are at an angle to one another; the induction planes use the “primed” coordinate system as shown in figure 12a. The nominal (default) detector coordinate system is identical to the Y plane’s coordinate system for which the $y$-axis is vertical in the upwards direction (toward the sky) and the $z$-axis is along the direction of the neutrino beam. The origin of the coordinate system is located at the center of the U plane’s upstream edge (the edge closest to the source of the neutrino beam). Based on the predefined coordinate for each wire plane, two angles define the direction of the track. As shown in figure 12b, $\theta_y$ is the angle between the track and the $y$-axis, and $\theta_{xz}$ is the angle between the projection onto the $x-z$ plane and the $z$-axis.

In estimating the full response, only tracks with $5^\circ < \theta_{xz} < 15^\circ$ are used in order to ensure that the impulse response of the drifting ionization electrons is well approximated, though cross-checks of the signal simulation against data are also performed at higher track angles. By repeating the measurement using a Monte Carlo simulation of cosmic ray tracks, it is determined that there is a Gaussian smearing of approximately 1 μs introduced by the method itself. The origin of this smearing is from the alignment of signals in time across many different waveforms, which introduces a slight broadening to the resulting average response. While delta rays from the cosmic tracks are not excluded in the analysis, this Monte Carlo study finds that the impact from delta rays to the broadening of the signal is negligible. When making comparisons of the full response between data and simulation, the 1 μs Gaussian smearing is applied to the simulation.

### 3.2 Response in normal region

By applying the methodology discussed earlier in this section, the full response in the normal region is extracted from data and compared to the convolution of the wire field response simulated by Garfield with a parametrization of the electronics response function for a given peaking time (2.2 μs in this case to match the effective peaking time seen in data). The comparison of the full response for data to that for the simulation in the normal region is illustrated in figure 13.
(a) Coordinates for collection and induction planes. The $y'$ ($z'$) axis is rotated by 60° around $x$ axis from $y$ ($z$) axis.

(b) Definition of two angles, $\theta_{xz}$ and $\theta_y$, which define the direction of a track in the TPC.

Figure 12. Geometric coordinates (a) and angles for description of track topology (b).

The absolute normalization of the full responses is arbitrary and only fixed relative to the integral of the Y plane response, which is set to unity.

In general there is quite good agreement between data and simulation, though there are a few differences worth pointing out. First of all, the U plane “front porch” or the slow-rising feature of the waveform in front of (at lower values of time) the primary “dip” exhibits some disagreement between data and simulation. This feature is due to the absence of a shield plane in front of the U plane in MicroBooNE, allowing charge further away from the anode plane to produce a signal on the unshielded U plane wires during drift (if there were a shield plane in front of the U plane, the U plane and V plane wire field responses would look very similar). The disagreement between data and simulation can be attributed to the finite number of wires that are adjacent to the wire closest to the drifting ionization electrons included in the Garfield simulation of the wire field.
Figure 13. Data/simulation comparison of the full response for the different wire planes in the normal region of the MicroBooNE TPC. Simulation is shown in black, while the colored curves represent the full response extracted from data. In general there is good agreement between data and simulation, but differences are observed between data and simulation in the “front porch” of the U plane response. This discrepancy is reduced by adding more wires in the wire field response simulation (see text for more details).
Figure 14. Comparison of U plane full response (electronics response convolved with wire field response) simulation to data, using a variable number of wires adjacent to the wire closest to the ionization electrons in the Garfield simulation: ±3, ±10, and ±50. This comparison is made in the region of the TPC unimpaired by shorted channels (normal region). The “front porch” of the U plane wire field response becomes less pronounced when the simulation utilizes a greater number of wires, converging to what is seen in data. A small positive peak before the “dip” of the U plane response is observed in data and is not reproduced in the simulation when utilizing any number of wires. This discrepancy is likely due to signal fluctuations present in data that are not included in the simulation.

In addition to differences in the “front porch” feature of the U plane, in figure 13 we can see a small discrepancy in the width of the signals between data and simulation. This difference in signal width is most noticeable on the Y plane but can also be seen on the U and V planes. The difference can potentially be explained by 3D effects that are not included in the Garfield simulation, which assumes that the wires of the different wire planes are parallel to one another and neglects the complicated drift of the ionization electrons through the wire planes in three dimensions. A potential improvement in the future is to use a full 3D simulation of the wire field response that is currently computationally expensive and too imprecise to utilize at present. The development of a sufficiently precise 3D simulation of the wire field response is in progress.

As described in ref. [11], charge induction on wires adjacent to the wire that is closest to drifting ionization charge can lead to large contributions to the signals observed on TPC wires. This effect is more noticeable when an ionization track is at higher angle with respect to the anode plane (i.e. higher $\theta_{xz}$). In order to probe how well the simulation performs in reproducing this effect as seen in data, a similar comparison to that illustrated in figure 13 is shown in figure 15 using tracks with $40^\circ < \theta_{xz} < 50^\circ$. The comparison shown in figure 15 indicates that the simulation performs
Figure 15. Comparison of the signal simulation for the different wire planes to data using tracks with $40^\circ < \theta_{xz} < 50^\circ$, looking at the region of the TPC unimpaired by shorted channels (normal region). Simulation is shown in black, while the colored curves represent the full response extracted from data.
Figure 16. The trajectories of ionization electron drift as a demonstration of the simulation of the electric field near the wire planes using Garfield. The ionization drift lines associated with the shorted-U region (a) and shorted-Y region (b) are shown, with a comparison made to the normal region in both cases. A couple of features are worth noting. In the case of the shorted-U region, a fraction of the charge that would be collected by the Y plane is instead collected by the unresponsive U plane wires, leading to a decreased transparency of the ionization electrons as seen by the V plane and Y plane wires. In the case of the shorted-Y region, ionization electrons that are normally collected by the Y plane are instead collected on the V plane wires, leading to unipolar (as opposed to bipolar) signals on the V plane wires in this region of the detector.

3.3 Response in shorted-wire regions

The comparison of the full response in data and simulation made in section 3.2 is repeated for the shorted-U and shorted-Y regions. In these regions, while the electronics response remains the same as in the normal region, the wire field response differs dramatically, as shown in figure 16. This is due to the different drift paths and velocities that ionization electrons experience as they drift through the anode plane wires, as a result of the different bias voltages on the shorted channels. Given the lack of information available regarding the nature of the short, data must be utilized in order to best model the true detector condition.

The short mechanism in the detector is still not completely clear. The simplest explanation of the short is that one or more wires in the V plane physically touch a group of wires in the U plane or Y plane; however, a visual inspection of the wire planes (prior to filling the cryostat with argon) via the use of a camera inserted into the cryostat yielded no evidence of wires physically touching each other [26]. In principle, the bias voltage on the wires shorted by this cross-plane contact should
Figure 17. Data/simulation comparison of the full response for the different wire planes in the shorted-U region. Simulation is shown in black, while the colored curves represent the full response extracted from data using tracks with $5^\circ < \theta_{xz} < 15^\circ$. The simulation predicts a slightly broader response than that observed in data, potentially due to the simulation neglecting ionization electron drift in three dimensions (see section 3.2).

be reduced to the ground level because of the contact with the V plane, which is held at detector ground. However, the fact that the sensitive wire is connected to a pair of diodes with 1.8 V bias for the pre-amplifier and ground, respectively, provides a hint that the actual bias on the wires may be different from ground. An imperfect short may cause residual voltage on the shorted channel significantly different from ground, with the supply bias set at $-110$ V and $+230$ V for the U plane and Y plane, respectively. Thus, the bias voltages on the shorted channels are set to values different from ground in the Garfield simulation in such a way as to reproduce the field response function shape as seen in data. For the shorted-U region, it was found that setting the bias voltage to $-45$ V leads to good agreement between data and simulation. A bias voltage of $+20$ V on the Y plane wires in the shorted-Y region is necessary to best match the simulation to data.

The data/simulation comparison for the shorted-U region is made in figure 17, utilizing tracks with $5^\circ < \theta_{xz} < 15^\circ$. As for the case of the normal region, the absolute normalization is arbitrary.
Figure 18. Data/simulation comparison of the full response for the different wire planes in the shorted-Y region. Simulation is shown in black, while the colored curves represent the full response extracted from data using tracks with $5° < \theta_{xz} < 15°$. As in the normal region, a small discrepancy is observed between data and simulation in the “front porch” of the U plane response that can be reduced by adding more wires into the wire field response simulation.

and only fixed relative to the integral of the Y plane response in the normal region (both for data and simulation, independently), which is set to unity as mentioned previously. Most of the shape features observed in data are reproduced with the modified Garfield simulation. As mentioned above, the transparency to ionization drift in this region of the TPC is reduced due to some of the ionization electrons being collected by the U plane wires, leaving less charge to induce a signal on the V plane and to be collected by the Y plane. The U plane response is not shown in this region as the U plane wires are unresponsive due to the short between the U and V planes.

The same comparison is made for the shorted-Y region in figure 18, again using tracks with $5° < \theta_{xz} < 15°$. As expected, the U plane response is largely the same as in the normal region due to the short occurring between the V and Y planes in this region. However, the V plane full response is dramatically different, unipolar instead of bipolar in shape. This is because ionization electrons are being collected on the V plane wires due to the Y plane wires in this region seeing
a dramatically lower positive voltage (zero in the case of a complete short to ground), leading to a loss of transparency for ionization electrons drifting through the anode wire planes. The modified Garfield simulation can reproduce the features seen in data in this region, in particular the shape of the unipolar V plane response. The disagreement between data and simulation in the “front porch” of the U plane response is similar to that observed in the normal region, discussed in section 3.2. The Y plane response is not shown in figure 18 because the Y plane wires are unresponsive in this region.

3.4 Signal processing for shorted-wire regions

As described in the previous section, there are two regions in the MicroBooNE TPC where the bias voltages on the wire planes are distorted such that the ionization electron drift paths are modified. Such alterations change the shapes of induced current produced on wires, requiring modifications to the signal processing chain. This is necessary in order to accurately determine the ionization electron distributions arriving at the TPC wires in these parts of the detector.

For the shorted-U region, the change in bias voltage in the U plane wires leads to a reduced number of ionization electrons passing through the V plane and being collected by the Y plane. Although there is clearly a reduction in the strength of the induced current as shown in the full responses extracted from data, the change in the shape of the field response function is limited. Therefore, no modification is made to the standard signal processing chain for the V plane and Y plane wires going through this region of the detector. Once the 3D positions of the ionization charge are deduced, the charge reconstructed in the shorted-U region is scaled up in order to compensate for the lost ionization electrons that are instead collected by the unresponsive U plane wires.

For the shorted-Y region, the situation is much more complicated. As the collection plane wires lose their bias voltage, some of the drifting charge is collected by the V plane wires positioned in front of them. The collection of ionization charge on these V plane wires leads to the formation of unipolar signals as opposed to the bipolar signals that are nominally seen. Meanwhile, these V plane wires also span the normal region where the passage of the ionization electrons through the V plane would lead to bipolar signals being produced. Therefore, both unipolar and bipolar signals can simultaneously exist on the set of V planes wires that physically overlap the shorted Y plane wires in this region. The standard signal processing chain attempts to interpret the measured signals with bipolar response functions. In this case, the existence of unipolar signals leads to large distortions in the deconvolved signal. These distortions propagate into the downstream event reconstruction chain, and are not desired. To deal with these situations, a special signal processing method based on compressed sensing techniques [27–29] was developed. These techniques were originally introduced into LArTPC event reconstruction [30] to improve the speed of the Wire-Cell 3D imaging reconstruction [15]. The central idea of compressed sensing is that an underdetermined linear system can be correctly solved by making assumptions about the intrinsic properties of potential solutions, such as sparsity and nonnegativity.

In the normal region, the measured signal is a vector with \( n \) time bins. The ionization charge signal \( S_1 \) to be deduced can also be expressed as a vector with \( n \) time bins. As discussed in section 1.1, they are connected by a presumably known bipolar response \( R_1 \), which is an \( n \times n \) matrix:

\[
M = R_1 \cdot S_1.
\]
Given \( n \) measurements and \( n \) number of unknowns, the extraction of \( S_1 \) essentially requires an inversion of the \( R_1 \) matrix, which in practice is performed through a discrete Fourier transformation. However, in the case of \( V \) plane wires going through the shorted-Y region, there are two types of signals and responses:

\[
M = R_1 \cdot S_1 + R_2 \cdot S_2.
\]  

(3.2)

Here, \( S_2 \) represents the ionization electrons collected by the \( V \) plane wires with the unipolar field response, \( R_2 \). In this case, there are \( n \) measurements, but the number of unknowns is increased to \( 2n \), leading to an underdetermined system. As a result, there are in principle an infinite number of solutions \((S_1, S_2)\) that can explain the same measurement \( M \). Fortunately, eq. (3.2) can be solved with compressed sensing techniques [27–29] through the application of L1 regularization. The solution to eq. (3.2) is obtained through a minimization of

\[
\chi^2 = (M - R_1 \cdot S_1 - R_2 \cdot S_2)^2 + \lambda \cdot \sum_i (|S_{1i}| + |S_{2i}|).
\]  

(3.3)

Here, \(|S_{1i}|\) represents the absolute value of the \( i \)th element of the unknown vector \( S_1 \) and \( \lambda \) is the regularization strength, a parameter to be chosen before minimization. The minimization of eq. (3.3) can be achieved through a coordinate descent algorithm [31], which is an iterative procedure. The final charge observed on the \( V \) plane wires would be the summation of \( S_1 \) and \( S_2 \). Due to the computational limitation of the coordinate descent algorithm, the signal processing used for the \( V \) plane wires going through the shorted-Y region (also referred to as “L1SP”) is limited to a 1D deconvolution instead of the 2D deconvolution that is used elsewhere. In addition, L1SP is only applied to the signal ROI (region of interest) instead of the entire waveform to limit the number of unknowns. \( R_1 \) contains the average bipolar field response and \( R_2 \) contains the average unipolar field response, with the regularization strength chosen to achieve an optimal result. The deconvolved signal from the standard signal processing chain is only replaced by that from L1SP if the strength of \( S_2 \) (unipolar) is significantly large.

Figure 19 illustrates the improved performance of the L1SP technique using waveform data associated with a \( V \) plane wire that overlaps the shorted-Y region. Compared to the results of standard 2D signal processing chain, the L1SP technique shows clear advantages in dealing with signals from the shorted-Y region, allowing for unbiased charge estimation in this region of the detector. An event display is shown in figure 20, highlighting a region of the detector where there are ionization tracks produced in the shorted-Y region and looking at the deconvolved \( V \) plane wire signals with and without the utilization of L1SP for the signal processing. The long tails introduced by the standard 2D signal processing chain are clearly removed using the L1SP technique.

4 Cleaning and filtering of data events

Additional steps are required in order to apply the signal processing chain described in section 1 to actual data collected by the MicroBooNE LArTPC. Various features unique to events in data are accounted for in order to put data and simulation on equal footing. Ultimately, this care is taken to improve the accuracy of physics measurements performed using MicroBooNE data.

A number of features have been observed in data events since the beginning of data-taking with the detector. These features are either difficult to introduce into simulation, or are best handled by
Figure 19. Illustration of the utilization of L1 signal processing (L1SP). A raw waveform (blue) is shown along with a comparison of two different signal processing algorithms performed on the waveform: the standard 2D deconvolution (pink, dashed) and the L1SP deconvolution (red). The first signal in the raw waveform is clearly bipolar and corresponds to a cloud of ionization electrons produced in the normal region. The second signal in the raw waveform is close to a unipolar shape and corresponds to another cloud of ionization electrons produced in the shorted-Y region, which is largely collected by V plane wires in this region of the detector. The deconvolved results of L1SP, measured in units of 1000 electrons after combining every six time ticks of the waveform, is compared with that of the standard 2D signal processing (2DSP), which contains a large distortion due to the incompatible field response used in the deconvolution. Improvement in the L1SP case is observed.

Figure 20. Event display highlighting the shorted-Y region of the detector, looking at the V plane wires post-deconvolution. Noisy and unresponsive channels are shaded grey in this illustration. Shown is the result for the standard 2D signal processing treatment (a) as well as results for the L1SP technique (b) in units of electrons per 3 μs. The long tails introduced by the standard treatment are clearly removed when the L1SP technique is utilized for signal processing of the V plane waveform data in this region of the detector.
correcting the data events directly (as negligible residual bias is expected after the correction). For instance, removal of signals on the TPC wires introduced by capacitive coupling to the PMT system is described in section 4.1. Another correction that addresses the shifting of bits in the ADCs is next discussed in section 4.2. Finally, covered in section 4.3 is the identification of “burst-like” signals on the TPC wires introduced by either the firing of the purity monitors in the detector or capacitive coupling to activity on the cathode or field cage.

4.1 Removal of PMT-induced TPC signals

The PMTs serve as the light collection system for MicroBooNE, enabling the $t_0$-tagging of cosmics and triggering of neutrino beam events (see section 3.1). When charged particles (e.g. cosmic muons) travel through the detector in the region between the PMTs and the anode planes, a large signal on the PMTs due to scintillation light is expected. It has been observed that groups of TPC wires also pick up signals in coincidence with charged particles traveling close to the PMTs (near the anode plane). This pick-up comes through the capacitive coupling between the PMTs and the TPC wires and is triggered by the large currents produced in the PMTs upon the arrival of a large amount of scintillation light. This phenomenon has been observed in both the ICARUS [32] and LArIAT [33] LArTPC experiments, and has been observed in MicroBooNE data since the beginning of data-taking. The characteristics of the PMT-induced TPC signals include:

- A range of collection plane wires that are located in front of the fired PMT simultaneously have a large negative raw signal corresponding to a positive current, which is different from the typical negative current induced by ionization electrons.

- At the same time, a range of induction plane wires that are located in front of the fired PMT also see a coincident signal. The strength of this signal is suppressed compared to that of the collection plane wires due to the shielding of the induction plane wires by the collection plane wires.

- These signals in the induction and collection plane wires are truly coincident (i.e. occurring at the same time tick). This is different from the induced current from ionization electrons, as ionization electrons take several microseconds to travel from one wire plane to the next.

A special algorithm is utilized to find and remove these PMT-induced TPC signals, without compromising the real signals induced by ionization electrons. For collection plane wires, this is straightforward, as the PMT-induced TPC signals behave quite differently from the normal ionization signals, with negative instead of positive signals on the waveform. Therefore, the PMT-induced signals can be removed efficiently by searching for negative signals that are of high amplitude with respect to the electronics noise level. Once the PMT-induced signal is found on the collection plane, the induction plane waveforms at the same time tick are searched for a group of induction plane wire signals that exhibit the signatures of PMT-induced signals. These identified PMT-induced signals are then removed from the waveform. Figure 21 shows a collection plane event display before and after the removal of the PMT pick-up noise, after carrying out the deconvolution in both cases. The pick-up feature is largely removed via the use of the technique described above.
Figure 21. Event display of the collection plane (Y plane) before (a) and after (b) removal of PMT pick-up noise, after carrying out the deconvolution in both cases in units of electrons per 3 μs. The event display is zoomed-in to a small part of the detector where PMT-induced TPC signals are observed in the event. The horizontal blue and red streaks seen in the bottom of the left image, which are attributed to capacitive-coupling between a PMT and the TPC wires, are largely removed via a dedicated algorithm.

4.2 Correction of ADC bit shifts

It has been observed that periodically groups of TPC channels connected to common electronics would experience a change of state, the main feature of which is a multiplication of the ADC values associated with the channel waveforms by some constant factor. This factor was observed to be a multiple of two (e.g., two, four, eight, etc.), a feature that is equivalent to a bit-shifting of the ADC values with a loss of higher bits. The problem is ameliorated by power cycling the warm readout crates. While the cause of this issue is not yet completely understood, given that changes in the problem only occur at the beginning of data-taking (when the ADC electronics are initialized), it is speculated that it may be due to an occasional misconfiguration of the ADC test pattern used in the dynamic phase alignment [34] circuit of the field-programmable gate array (FPGA) that sets the correct order to read out the ADC data bits for a given channel. A visual depiction of the ADC bit-shifting issue is shown in figure 22.

In order to identify problematic channels affected with the ADC bit-shifting issue, a waveform-level algorithm is run on every channel in the noise-filtering step of event reconstruction. This algorithm finds the number of bits shifted in the ADC values of a channel waveform by determining the bit that fluctuates the most between adjacent time bins. This bit should be the true lowest bit as even small variations throughout the extent of the waveform, due to noise, may alter the value of this bit before it is shifted in the warm electronics. Each channel with a non-zero bit shift is categorized as a “bad” channel that is corrected by another algorithm described below. It is observed that a bad channel remains bit-shifted by the same number of bits for the entirety of a data-taking run (up to seven hours in length), with transitions only happening at run boundaries (when the warm electronics are reconfigured). After the first observation and characterization of this problem,
Figure 22. Visualization of the ADC bit shift issue and how it impacts ADC values in waveform data, as well as the basic concept of the correction method utilized to recover channel waveform data affected by the ADC bit shift issue. A small part of a waveform from a channel affected by the ADC bit shift issue is shown (a). Each ADC value on the waveform is shifted by the same number of bits, upward, leading to the loss of the highest $N$ bits for a $N$-bit shift. The lowest $N$ bits are replaced with new values, as illustrated in the table at the bottom of the figure. The corrected waveform is also shown (b). In addition, the correction of a single ADC value is illustrated (c); for a bit-shift of $N$ bits, the highest $12-N$ bits are shifted down $N$ places. The new highest $N$ bits are selected from a collection of lowest bits associated with each of the time bins on the given channel’s waveform before any correction is applied. These highest $N$ bits are chosen based on their compatibility with an interpolation/extrapolation using the two time bins before and after the waveform time bin of interest, enforcing the fact that a channel’s waveform should be relatively smooth moving from time bin to time bin. This procedure is repeated for every channel in the event.

improvements were made to the monitoring of the detector to allow for quick response to this issue by the MicroBooNE operations team. However, some earlier data-taking runs at MicroBooNE are impacted by the ADC bit shift issue and require special treatment.

A correction for each affected channel is attempted before carrying out extensive noise filtering and signal processing of the TPC channel waveforms. This correction is illustrated in figure 22. As the number of bits by which the ADC values of the channel waveform are shifted is known, the highest bits of the observed waveform can be shifted back down to partially restore the original waveform. After this operation, one must still determine the highest bits of the ADC values of the channel waveform that were lost when the original waveform was bit-shifted. Empirically it was found that these highest bits were almost always shifted into the lowest bits of the waveform ADC values. In other words, the lowest bits of the observed waveforms associated with bad channels are not random and can be corrected. However, the lowest bits of the waveform are not always from the same time bin that the bits were originally associated with. In order to pick which set of bits to
assign to the highest bits of the corrected waveform (from the collection of lowest bits associated with each of the time bins on a given channel’s waveform before any correction is applied), a smoothness criterion is employed. This criterion, utilizing a combination of interpolation between and extrapolation across neighboring time bins (two in each direction), makes use of the fact that a channel’s waveform should be relatively smooth moving from time bin to time bin. An example of this correction on a subset of an affected waveform is shown in figure 22.

In order to validate the correction described above, a metric that makes use of “good” channels (channels with waveforms unaffected by the ADC bit shift issue) physically neighboring the bad channels was developed in order to quantify how well the correction method performs on actual data. This validation technique relies on the fact that minimum-ionizing particles (cosmic rays) will produce ionization signals that are similar in magnitude for neighboring wires. Therefore, we expect the “fixed” channels (bad channels that have been corrected via the method described above) to yield measured ionization signals that are similar in amplitude to a “reference” channel, the closest good channel in the TPC. The construction of this metric is described visually in figure 23. The metric utilizes the maximum ADC value on the waveform relative to the median ADC value on both the reference channel and the bad channel, taking the difference between the two. This is done both before and after the correction is applied to the bad channel. By repeating the metric calculation using a good channel instead of a bad channel, the distribution for a “perfect correction” is recovered.

\[
\text{Metric: } \left| \left( \text{Max}(R) - \text{Median}(R) \right) - \left( \text{Max}(K) - \text{Median}(K) \right) \right|
\]

where \( K = \{G, B, F\} \)

![Figure 23. Visualization of the construction of the ADC bit shift metric, which is utilized to validate the correction of the ADC bit shift issue. The metric is constructed by taking the maximum ADC value on the waveform relative to the median ADC value on both the reference channel (chosen to be the closest good channel preceding a string of bit-shifted channels) and the bad channel, taking the difference between the two. The same metric can be constructed using a fixed channel (bad channel after the correction procedure has been carried out) or good channel instead of a bad channel. By comparing the metric distribution constructed using the fixed (corrected) channels and the metric distribution constructed using good channels that are known to be unaffected by the ADC bit shift issue, one can validate the performance of the correction procedure using data.](image-url)
The validation procedure outlined in figure 23 is carried out for bad, fixed, and good channels. A comparison of the ADC bit shift metric between good and bad channels is shown in (a) and (c), while a comparison of the ADC bit shift metric between good and fixed channels is illustrated in (b) and (d). Both full distributions of the ADC bit shift metric values, (a) and (b), and zoomed-in distributions, (c) and (d), are shown. After the application of the correction method discussed previously, the metric distribution shows excellent agreement with respect to a “perfect correction” that is found in data using the comparison between good channels and a reference channel. In the axis labels, “R” refers to a reference channel and “K” can be either “G” (a good channel), “B” (a bad channel), or “F” (a fixed bad channel).

Figure 24. Results of performing the ADC bit shift correction on affected channels in MicroBooNE data. The distribution of the ADC bit shift correction metric using bad channels, fixed channels, and good channels along with a reference channel is shown in figure 24. Several hundred events from an early data-taking run that was significantly impacted by the ADC bit shift issue is utilized in this validation, including all affected channels in the event. It is seen that the correction performs well, with the distribution of the metric using fixed channels nearly matching the distribution using good channels.

### 4.3 Handling purity-monitor-induced burst noise and cathode burst events

Throughout data-taking with the MicroBooNE detector, a number of external noise features were observed in TPC data. In addition to the continuous external noise features described in ref. [13], there have been burst-like signals occurring on many different TPC channels at the same measured drift time that are not associated to particle-induced ionization. At present this burst-like noise has been categorized into two distinct types of signals. The first one is observed only when the liquid argon purity monitors [8] are functioning and is associated with a large amount of light collection in the PMTs; this noise is most intense on the collection plane (Y plane), the wire plane closest to the PMT system. The second one is most intense on the first induction plane (U plane), the wire plane closest to the cathode, and is not associated with an excess of light seen by the PMT system.
The fact that channels from all three planes share the same electronics, but the signals are seen predominantly on the U and Y planes, respectively, for the two classes of noise, points to a source of noise external to the electronics. Given the locations of the U and Y wire planes in the cryostat, and correlations with the PMT and purity monitor activity, it is currently believed that the first type of noise is associated with the firing of the purity monitors (due to a light leak from the purity monitor flash lamps) and the second is due to capacitive coupling between voltage fluctuations in the TPC cathode plane and the U plane wires. For this reason, the former issue is referred to as “purity-monitor-induced burst noise” and the latter issue is referred to as “cathode burst noise” in the discussion below. An example of the purity-monitor-induced burst noise and cathode burst noise is shown in figure 25 and figure 26, respectively.

As discussed above, the purity-monitor-induced burst noise is identifiable by both the striking signature in the TPC waveforms (most dramatic on the Y plane) and the large amount of light seen in the PMTs in the same data event. While the TPC noise feature is isochronous, the signature seen on different wires are not completely coherent across the entire wire plane, and so this noise cannot be completely removed using coherent-noise removal techniques [13]. These coherent-noise removal techniques are also designed in such a way as to prevent the removal of isochronous signals on the TPC wires due to tracks traveling nearly parallel to the wire planes, further complicating the removal of burst-like noise features. Additionally, the excess of light seen in the PMTs due to the presence of the burst noise provides another complication as this light makes it more difficult

Figure 25. Example of an event in MicroBooNE data that is impacted by the purity-monitor-induced burst noise issue. Shown is an event display for the waveform data associated with Y plane wires after coherent noise filtering is carried out in units of electrons per 3 μs. Noisy and unresponsive channels are shaded grey in this illustration. One distinctive feature of the purity-monitor-induced burst noise is a strong isochronous signal across all wires that is most severe on the Y plane (as seen by the horizontal band in the display). Another feature of this specific type of burst noise is the large amount of light observed in the PMTs in the same event (discussed later). The noise feature is not removed from data by the coherent noise filtering procedure that is applied to raw data.
Figure 26. Example of an event in MicroBooNE data that is impacted by the cathode burst noise issue. Shown is an event display for the waveform data associated with the U plane wires after coherent noise filtering is applied in units of electrons per 3 μs. Noisy and unresponsive channels are shaded grey in this illustration. The distinctive feature of the cathode burst noise is a strong isochronous signal across all wires that is most severe on the U plane, which is seen as a band at the top of the event display. The coherent noise filtering step does not remove this noise feature from data.

to reconstruct the $t_0$ of cosmic activity in the event. As a result, there is no simple algorithm to completely remove the burst noise from TPC and PMT data; instead, an attempt is made to filter out these events from consideration in MicroBooNE physics analyses. This only needs to be done for data-taking runs prior to April 2016, after which the purity monitors were permanently turned off in order to mitigate this noise source. Prior to the purity monitors being turned off, the fraction of events impacted by this type of burst noise depends on the relative firing rate of the purity monitors, leading to as many as 10% of events impacted by the noise in a given data-taking run (lasting several hours). However, most runs taken while the purity monitors were online see less than 0.5% of events impacted by the purity-monitor-induced burst noise.

It was found empirically that the high amount of light produced in these events, collected by the PMTs, is enough to tag the purity-monitor-induced burst noise events. This was determined by comparing two data-taking runs: one before the purity monitors were turned off, and one after. In both runs, the number of “flashes” in the event is measured for all of the events in each of the two data-taking runs, where a “flash” is defined as light observed in multiple PMTs at roughly the same time (in a window of tens of nanoseconds). In order to enhance the separation of the two populations (normal events and purity-monitor-induced burst noise events), a metric was developed (“nflash500”) which is the maximum sum of the number of flashes in the event in a 500 μs window, scanning the entire event in time. By placing a cut of $n_{\text{flash500}} < 40$, we are able to effectively remove all events from the dataset that contain the purity-monitor-induced burst noise feature.
For the cathode burst noise, there is no enhancement of light production in the event and so one must rely strictly on the TPC data in order to identify events with the noise feature present. A common feature of this particular category of burst noise is a large signal, mostly, but not completely, coherent across all U plane channels, that is very extended in time (on the order of a millisecond). As a result, the burst noise signal has a very low-frequency component (< 2 kHz) that is not present in the bipolar induction plane signals from ionization electrons traveling by the U plane. Furthermore, other coherent noises do not have power in the relevant frequency range. The presence of this power in the low-frequency range allows us to identify and remove these events from our dataset. A metric (“fftsum2”) for this tagging was constructed by integrating the first 2 kHz of the power spectrum of the U plane after adding together all channel waveforms associated with that plane, coherently in time. A cut of fftsum2 < 50000 is applied in order to mitigate the cathode burst noise events. In February 2017, it was determined that the cause of this specific type of burst noise was due to an intermittent connection between the cathode HV feedthrough and the cathode plane. After readjusting the cathode HV feedthrough, the noise was no longer observed in significant amounts in MicroBooNE data.

Two-dimensional plots showing the distribution of both nflash500 and fftsum2 for two data-taking runs (one before the purity monitors were turned off and one after they were turned off) are shown in figure 27. It was noticed in the run after the purity monitors were turned off that the cathode burst noise was reduced.

![Figure 27](image)

(a) Values of burst noise metrics in a noisy run.

(b) Values of burst noise metrics in a quiet run.

**Figure 27.** Two-dimensional plot showing both the nflash500 and fftsum2 distributions for two data-taking runs, one that was empirically observed to have higher levels of both cathode burst and purity-monitor-induced burst events (a), and one where very little of either burst noise was observed (b). The cut boundaries in both distributions used to remove the problematic events are shown in red. The cuts chosen effectively remove the problematic events while leaving “normal” events in the sample.
burst noise also reduced greatly in magnitude, though this noise feature has been seen in event data after the purity monitors were turned off as well. Similar to the case of the purity-monitor-induced burst noise, the cathode burst noise impacts as many as 20% of the events in some data runs, although these runs are relatively rare; in most runs, less than 0.5% of events are significantly impacted.

The cuts on nflash500 and fftsum2 described above largely remove the purity-monitor-induced burst noise and cathode burst noise events. However, the cathode burst noise events are not completely removed by this cut. It is difficult to remove all cases of this noise because some of the cathode burst noise signals are very small in magnitude; because of this, it is also difficult to quantify the fraction of events that remain impacted by the cathode burst noise after application of the cuts. Near the fftsum2 cut boundary, they are difficult to identify visually on individual waveforms, an indicator that the cut is sufficient in removing any noise that would be troublesome for downstream reconstruction of the event. It is worth noting that a very small fraction of “good” events (less than 0.1%) are removed by the cuts. These events were studied and determined to be associated with unusually large cosmogenic showers that can be removed in the study of lower-energy (< 2 GeV) neutrino interaction events, such as those produced by the BNB in the MicroBooNE detector.

5 Evaluation of signal processing performance on MicroBooNE data

With the calibration of the electronics response in data, the tuning of the wire field response simulation to match data, the handling of the shorted wire regions in the signal processing chain for data events, and the cleaning or removal of certain noise events unique to data, the data collected by the MicroBooNE detector demonstrates much better agreement with simulated data. With these conditions met, it is possible to evaluate the performance of the new signal processing chain (including the 2D deconvolution) using MicroBooNE data (hereafter, “deconvolution” will refer to the 2D deconvolution, unless otherwise specified).

Two general metrics are used to evaluate the performance of signal processing on data events: the qualitative improvement in image quality after the signal deconvolution, discussed in section 5.1, and the quantitative improvement in cross-plane charge matching after the full signal processing chain, highlighted in section 5.2. The former is simple to interpret; better image quality after deconvolution is expected to improve the pattern recognition stage of neutrino interaction reconstruction. The latter is helpful in quantifying improvements in the charge extraction step of event reconstruction. It is also a necessary demonstration for the application of novel LArTPC event reconstruction techniques like Wire-Cell tomographic reconstruction [15], which uses local charge correlations across two or more wire planes of a LArTPC in order to create a three-dimensional distribution of points. Each point in the distribution has an associated charge that is estimated from the waveform data associated with overlapping wires from different planes that share signals from the same drifting ionization charge. This approach allows pattern recognition on three-dimensional images, as opposed to first doing pattern recognition on the two-dimensional images from each wire plane, followed by matching across views. This method is expected to have advantages in charge clustering and potentially neutrino interaction vertex-finding, as it is performed in three dimensions rather than two. In order to realize this reconstruction technique, it must be shown that charge can be matched across the three wire planes using events in data, which includes the demonstration of the precise extraction of charge information from the induction planes of a LArTPC.
5.1 Event displays

In order to illustrate the qualitative improvement in imaging with the updated signal processing chain, a variety of different events have been studied in data, making comparisons before and after the deconvolution step. In this section a number of event displays are shown and discussed in order to demonstrate this improvement in terms of higher image quality. By high image quality, we mean that charge deposition in the detector, processed through the signal processing chain, lead to clearly recognizable patterns that correspond to the production of physical objects in the detector, e.g. charged particle tracks and electromagnetic showers (from electrons or photons).

The event displays shown in this section are of localized sections of the entire TPC, without restriction to any particular section of the detector (as was the case for the studies completed in section 3, where the ionization charge depositions were restricted to be between 2 cm and 10 cm away from the anode plane in order to minimize effects of diffusion). In all cases, the data was recorded while the neutrino beam was on under normal MicroBooNE operating conditions. The events were selected to demonstrate the impact of applying signal processing algorithms to neutrino interaction candidates.

Figure 28 illustrates a busy neutrino interaction candidate event with several tracks produced in the final-state. Shown are the raw, unprocessed image (“Raw”), the image after software noise filtering is carried out (“After Noise Filtering”) and finally the image after the full deconvolution is

![Event Displays](image)

**Figure 28.** An example neutrino candidate event display from MicroBooNE data (event 41075, run 3493) showing a U plane view. (a) The raw waveform image in units of average baseline subtracted ADC scaled by 250 per 3 μs. (b) The image after software noise-filtering in units of average baseline subtracted ADC scaled by 250 per 3 μs. (c) The image after 2D deconvolution in units of electrons per 3 μs. Prolonged signals associated with near-vertical tracks, such as the one at the top left of each event display window, are recovered after the deconvolution step. Additionally, the image quality near the neutrino interaction vertex improves after the 2D deconvolution, which is expected to lead to improvements in the pattern recognition.
Figure 29. An example neutrino candidate event display from MicroBooNE data (event 41075, run 3493) showing a Y plane view. (a) The raw image in units of average baseline subtracted ADC scaled by 250 per 3 μs. (b) The image after software noise-filtering in units of average baseline subtracted ADC scaled by 250 per 3 μs. (c) The image after the 2D deconvolution in units of electrons per 3 μs are shown. The impact from the deconvolution is minimal in the case of the Y plane, but minor improvement is observed; most notably, faint vertical lines near the neutrino interaction vertex are removed. The improvement in this case is largely a result of the calibration of the electronics response.

Another event display, showcasing the same neutrino interaction shown in figure 28 but this time in the Y plane view, is illustrated in figure 29. For the collection plane, the signal deconvolution has a much smaller impact compared to that on the induction plane views. However, in figure 29 one can see a small improvement in the image quality near the neutrino interaction vertex and close to the particle tracks after the deconvolution. In particular, the faint vertical lines near the vertex (and along some of the tracks) are removed. This is largely due to the calibration of the electronics response that is discussed in section 2. Although the difference is not dramatic visually, the faint vertical lines can result in fake tracks, negatively impacting the pattern recognition.

Figure 30 shows an event display of the U plane from a different event, and shows a similar improvement after the deconvolution. The delta ray (low-energy knock-out electron) associated with the longer muon track at the bottom of the display comes into focus, and the short vertical track pointing to the vertex is recovered. This short track may be associated with a proton that was knocked out of an argon nucleus by a neutron associated with the neutrino interaction. Accurately identifying tracks near neutrino interaction vertices is important for neutrino event reconstruction at MicroBooNE as well as at DUNE [35].
Figure 30. An example neutrino candidate event display from MicroBooNE data (event 53223, run 3469) showing a U plane view. The image after software noise-filtering in units of average baseline subtracted ADC scaled by 250 per 3 μs (a) and the image after the 2D deconvolution in units of electrons per 3 μs (b) are shown. The delta ray in the bottom right part of each event display window becomes more crisp after the deconvolution, enabling more accurate pattern recognition. Additionally, a short track near the neutrino interaction event is recovered after the deconvolution.

Shown in figure 31 is yet another event display corresponding to the event showcased in figure 30, this time for the V plane view. As for the case of the U plane for the same event, the delta rays associated with the longer muon track at the bottom of the display and the neutrino interaction vertex become more easily identifiable by eye in the image after the deconvolution, and the short track near the neutrino interaction vertex is recovered. However, one also notices some blurring of the event image near the neutrino interaction vertex. This is in part due to a deficiency in the ROI (region of interest) finding for this particular event (see section 1.1). Future improvements in the ROI finding algorithm as well as a better modeling of the wire field response will allow for further improved image quality.
Figure 31. An example neutrino candidate event display from MicroBooNE data (event 53223, run 3469) showing a V plane view. The image after software noise-filtering in units of average baseline subtracted ADC scaled by 250 per 3 μs (a) and the image after the 2D deconvolution in units of electrons per 3 μs (b) are shown. The delta rays in the bottom right of each event display become more easily identifiable in the resultant image after the deconvolution; however, there is slight blurring of the neutrino interaction vertex due to the deconvolution step. This is one example of where additional improvements in the signal processing chain can help improve image quality of neutrino interaction events.

5.2 Demonstration of cross-plane charge matching

It has been shown that with the improved signal processing chain, drifting ionization charge can be successfully extracted from the induction wire planes where the intrinsic field response function has a bipolar shape. In this section, it is shown quantitatively that charge matching across the three wire planes (the unbiased estimation of amount of ionization charge from the same drifting electron cloud when comparing signals from the different wire planes) has been achieved in MicroBooNE data.

Figure 32 shows the extracted deconvolved ionization charge distribution for a set of selected “point sources” from small, isolated ionization deposits in the TPC such as beta decay electrons from $^{39}$Ar or other, higher-energy radiological sources in the liquid argon. The distributions shown are obtained by adding together all deconvolved waveforms of a given plane associated with a point source in the detector that is identified by eye. In general, the charge distribution as a function of time from different wire planes are reasonably consistent with each other. At the same time, some discrepancies do exist in these comparisons. Some differences can be attributed to the inherent (or residual excess) electronics noise. In the bottom left image of figure 32, the V plane wires make
use of a bigger signal ROI than is necessary, leading to a reduction of reconstructed charge in the first peak. Future improvements to the ROI-finding algorithm, described in section 1.1, can help ameliorate this problem. Besides this, bias in the charge extraction can result from deviations of the local wire geometry from the ideal case (due to a slight sagging of the wires, for example) as well as from the use of inaccurate field response functions in the deconvolution, among other things. These discrepancies can be minimized with additional studies that are beyond the scope of this article.

Figure 32. Individual deconvolved ionization charge distributions for a set of selected point sources are shown to demonstrate charge matching across the three wire planes. These distributions are obtained by summing the deconvolved waveforms associated with the extent of charge deposition from each point source (the sum over the range of wires where a signal from the point source can be detected), independently for each plane. The numbers in the legends show the integral of the charge distributions for each plane, with units in thousands of electrons.

Figure 33 compares the per-plane extracted deconvolved ionization charge distributions for two example “line sources” produced from cosmic muons passing through the TPC. As for the point source case described above, these distributions are obtained by summing together the deconvolved waveforms associated with each line source on a per-plane basis. Reasonable agreement in charge matching across the three planes is demonstrated, as in the point source case. The remaining differences can be explained by the inherent electronics noise, which is amplified in the deconvolution process, as well as inaccuracies in the field response functions obtained from Garfield. The features in these distributions are due to large signal fluctuations in the deposited energy along the muon track, including delta rays.

In addition to individual point and line sources, the average deconvolved signal shape associated with tracks of different angles in the detector is shown in figure 34. Results are shown both for the case of the 1D deconvolution and the improved 2D deconvolution in the left and right columns of figure 34, respectively. It is clear from this comparison that the matching of signal shapes across planes is significantly improved when moving from the 1D to the 2D deconvolution. In the case of the 2D deconvolution, the average deconvolved signal shape is very close to a Gaussian, which is expected, because a Gaussian filter is used in the signal processing (see section 1.1).
Figure 33. Individual deconvolved ionization charge distribution for two example line sources, associated with cosmic muons passing through the detector, are shown to demonstrate charge matching across the three wire planes. These distributions are obtained by summing the deconvolved waveforms associated with the extent of charge deposition from each line source (the sum over the range of wires where a signal from the line source can be detected), independently for each plane. The numbers in the legends show the integral of the charge distributions for each plane, with units in thousands of electrons.

Improvement in moving from the 1D to the 2D deconvolution is most noticeable for the tracks that are oriented more in the drift direction (higher angles in figure 34), as the wire field responses used in the case of the 1D deconvolution were derived from tracks nearly parallel to the anode plane. The 2D deconvolution is able to account for the charge induction profile of drifting charge on neighboring wires for all track angles.

To further quantify the level of charge matching across the three wire planes, a more detailed comparison is made utilizing all of the ionization charge collected in a readout event. First, the entire drift time window is divided into many different independent time slices. Each time slice consists of four time ticks, corresponding to 2 μs, which is about 2.2 mm of drift in liquid argon. In each time slice, adjacent intraplane wires with observed signals are grouped together to form a “merged” wire. These merged wires are then compared with merged wires from the other two wire planes in order to search for signal activity (referred to hereafter as a “blob”) in the transverse \((y - z)\) plane [15]. As the entire drift time window is used, the signal activity may originate from any distance away from the anode plane (unlike in the studies of section 3). The top row of figure 35 shows the extracted charge from the induction wire planes (U plane on left, V plane on right) for a blob versus the extracted charge from the collection wire plane (Y plane) for the same blob, utilizing the 2D deconvolution for signal processing. Good matching of charge across the wire planes is generally observed in these distributions. For the U plane vs. Y plane comparison, there is a small group of points which has a reduced U plane charge in comparison to that extracted from the Y plane. This is due to prolonged signals associated with tracks nearly orthogonal to the wire planes, which are better imaged with the 2D deconvolution, but still experience a downward bias in reconstructed charge magnitude. Also, for the V plane vs. Y plane comparison, there is a small group of points which have a much enhanced V plane charge with respect to that extracted from
Figure 34. Comparison of average deconvolved signal shapes (both 1D and 2D deconvolution) on the U, V, and Y planes using cosmic muons near the anode plane at a variety of angles in data. $\theta_{xz}$ is the angle of the track between the drift direction and the direction in the wire plane ($y$-$z$ plane) perpendicular to the wire orientation (see figure 12b). All distributions are normalized to unity.
Figure 35. Using a sample of cosmic muons in data, the extracted charge from the induction U plane and induction V plane for a “blob” (matched set of wires with signal across the three planes of the LArTPC) is plotted against the charge from the collection Y plane (a). The diagonal lines go through the origin and have unity slope. Good matching of charge between the induction planes and collection plane is observed. The projection of these scatterplots along the unity-slope line are also shown for the U plane and V plane, normalizing to the charge observed by the Y plane in both cases (b).

The spread about the unity-slope lines in the top row of figure 35 is dominated by the charge resolution in the induction wire planes. This is illustrated more clearly in the bottom row of the same figure, where the spread about the unity-slope line is profiled for both induction planes, normalizing to the charge observed by the Y plane. By fitting the core of each distribution to a Gaussian and assuming perfect knowledge of charge using the Y plane, it is determined that the charge resolution of the blobs is roughly 11% using the U plane signals and roughly 14% using the V plane signals. However, this is not equivalent to the per-wire charge resolution, as the blobs used in figure 35 are...
formed by merging signals (and noise) across neighboring wires and adjacent waveform time bins. The charge resolution can in principle decrease (increase) by merging more (less) information in this way. Furthermore, the charge resolution for the induction planes depends heavily on the signal window size. This varies greatly across different particle topologies observed in the TPC readout. As a result, it is not appropriate to interpret the above charge resolution numbers as applicable to all signals observed in data for a given wire plane. There is also a small bias in average extracted charge between the V plane and Y plane (roughly $-4\%$, obtained via the Gaussian fit), as presented in figure 35. This is likely due to an imperfect estimation of the V plane wire field response (discussed in section 3).

While the charge resolution for the collection plane is very similar between data and simulation, the charge resolution for the induction planes is about a factor of two worse in data than that predicted by studies utilizing simulated events [11]. There are several potential reasons for this. First, in the full TPC simulation, the field response functions are also used in the TPC signal deconvolution. These field response functions are based on a two-dimensional Garfield simulation, and are slightly different from those realized by the MicroBooNE three-dimensional wire geometry. Additionally, the local wire geometry may deviate from the ideal case, where the wire pitch and gap between wire planes are exactly 3 mm. Such deviations would lead to differences in the field response observed in data. Finally, the software noise filtering that is required to remove excess noise on TPC channels (before the hardware was modified [13]) introduces some bias in the raw waveform, which can be propagated to the deconvolved waveform. This may not be sufficiently taken into account in the simulation, in which only the inherent electronics noise is simulated. The detailed investigation of these differences is beyond the scope of this article.

In another demonstration of the cross-plane charge matching capability of the improved signal processing chain in MicroBooNE, the cosmic muon $dQ/dx$ distribution obtained using cosmic muon tracks in MicroBooNE data is shown in figure 36 for both the original 1D and the new 2D deconvolution. A wide variety of track angles is utilized in this sample. In the case of the 1D deconvolution, the three planes do not align in terms of the estimated values of $dQ/dx$ using each plane individually. In the 2D case, the agreement among the planes is better, especially for the most probable value of energy loss (MPV); the spread in the values of MPV across planes reduces from 15% in the 1D case to 4% in the 2D case as determined from fits to a Landau convolved with a Gaussian. The induction planes experience a lower performance in terms of $dQ/dx$ resolution than the collection plane, despite there being little bias across the three planes. This is in agreement with expectations from charge resolution studies carried out in ref. [11] using simulated events. Furthermore, these $dQ/dx$ distributions sample over a variety of different residual ranges (distance from the stopping point of the particle), the only requirement being that the cosmic muon pierces the anode plane and only the charge between 2 cm and 10 cm from the anode plane is included in the $dQ/dx$ measurement. This reduces the impact from various detector effects, such as the finite electron lifetime and the diffusion of the drifting electrons in the liquid argon. This leads to additional broadening, similar for all three planes, in comparison to the distribution one might see when looking at a fixed distance from the end of a cosmic muon stopping in the detector.

In summary, the successful extraction of drifting ionization charge from induction wire planes and cross-plane charge matching in a LArTPC has been demonstrated with MicroBooNE data. The reconstructed charges in the three wire planes after deconvolution agree reasonably well for both individual point and line charge sources as well as average signal shapes in MicroBooNE data.
Figure 36. Cosmic muon $dQ/dx$ as estimated from each plane (separately) using cosmic muon tracks in data. Results are shown both for the case of the 1D (a) and the improved 2D (b) deconvolution. Charge estimation is done by integrating across each waveform, while the track angle used in the $dQ/dx$ calculation is obtained from the output tracks created by the Pandora pattern recognition [23]. Only charge deposited between 2 cm and 10 cm from the anode plane is included in the $dQ/dx$ calculation. Tracks with a wide range of angles are used in the sample.
Further studies are needed in order to improve the level of agreement. Of primary importance is to improve the calculated field response function to better match the field response observed in data, as discussed in section 3.

6 Summary

A robust signal processing chain was developed for MicroBooNE in order to extract charge information from signals on all three TPC wire planes with minimal bias and good charge resolution [11]. Calibration of the electronics chain and validation of the response simulation using data events was necessary to accomplish this. The conclusion from these studies is that the signal shapes in data and simulation experience much better agreement after calibration. A number of steps were required to achieve this result, starting with additional cleaning of data events, and identifying and removing certain relatively rare events compromised by excess noise signals. We have demonstrated an improved image quality and unbiased charge matching across the three wire planes of the MicroBooNE LArTPC. We have also presented the first quantitative evaluation of extracted charge measurements using the 2D deconvolution algorithm in LArTPC data. This detailed understanding of the MicroBooNE TPC response and the performance of signal processing algorithms on data is crucial for the development of physics analyses and evaluation of systematic uncertainties, as well as for the design, operation, and calibration of future LArTPC experiments.

This article has presented the precise extraction of ionization charge from induction plane wires in a single-phase LArTPC, which is expected to improve event reconstruction and provide more accurate charge measurements. In addition, accurate charge matching across wire planes has been demonstrated for the first time, allowing reconstruction of three-dimensional charge distributions as in Wire-Cell tomographic reconstruction [15]. This is expected to improve the reconstruction of neutrino interactions and cosmic-ray muons.

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