

Analysis Note of the Experiment

Supplementary note on diffractive analyses of 2015 proton-proton data:

- [1] Measurement of Central Exclusive Production of $h\bar{h}$ pairs ($h = \pi, K, p$) with Roman Pot detectors in diffractive proton-proton interactions at $\sqrt{s} = 200$ GeV
- [2] Measurement of particle production with Roman Pot detectors in diffractive proton-proton interactions at $\sqrt{s} = 200$ GeV

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Abstract

In this note we present supplementary material related to analyses of diffractive processes based on 2015 data from proton-proton collisions at $\sqrt{s} = 200$ GeV. This dataset was collected with newly installed Roman Pot detectors in Phase II* configuration which ensured efficient triggering and measuring diffractively scattered protons.

We focus on issues which are common in aforementioned analyses: calculation of efficiencies, corrections to Monte Carlo simulations and derivation of systematic uncertainties of our measurements.

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List of contributions

Leszek Adamczyk	Analysis coordination/supervision, production of picoDST, production of embedded MC samples
Lukasz Fulek ¹	pp2pp Geant4 simulation (user interface, beamline, magnet lattice, B -field), TPC efficiency calculation, TOF efficiency calculation, matching between true-level particles and reconstructed TPC tracks, energy loss corrections, pile-up-related systematics to TPC and TOF efficiency
Rafal Sikora ²	pp2pp Geant4 simulation (user interface, RP detectors, digitization and reconstruction, embedding), dE/dx adjustment, TPC track resolution adjustment, pre-TPC dead material study, corrections to TOF efficiency and related systematics, RP efficiency systematics, list of bad runs

¹ - contact editor for Chapter: 2, 3 (Sec. 3.1-3.2.2), 5, 10 (Sec. 10.1-10.2.1)

² - contact editor for Chapter: 1, 2, 3 (Sec. 3.2.3-3.3), 4, 6, 7, 8, 9, 10 (Sec. 10.2.2)

Change log

?? Dec 2018 ver. 1.0 Initial revision

1. Bad run list

Diffractive analyses [1, 2] were performed with the use of data from runs with completion status “Successful” in the STAR run log [3]. During analysis of these runs no problems with the reconstructed data were found. The distributions of the TPC track-related quantities and performance of remaining subsystems used in analyses did not indicate issues preventing this data from inclusion to physics analyses.

We omitted from analyses data taken with Roman Pot detectors far from their typical positions. By the position we understand the vertical distance between the nominal beam trajectory and the closest edge of the Silicon Strip Detector housed inside Roman Pot. These positions were first measured during dedicated survey [4] and later precisely calculated for all runs and for all detectors using elastic proton-proton scattering events [5, 6].

Below we show histogram (Fig. 1.1a) and graph (Fig. 1.1b) of the beam-detector distance of all Roman Pots during runs with active RP triggers¹. Based on these plots we set the limit on the actual position of the detector with respect to the beam to $|y_{\text{thr}}| = 34 \text{ mm}$. If any of 8 detectors was further from the beam than the threshold ($|y| > y_{\text{thr}}$) the run was omitted from analysis. The benefit of that selection was reduction of the systematic effects related to low/asymmetric acceptance of detectors when they are at far/significantly differing distances from the nominal beam trajectory.

In the Fig. 1.1 one can notice that there were a few runs (16106026-16106033, fill 18915) with the closer-than-typical distance of detectors to the beamline. These runs correspond to a period dedicated for a study of elastic proton-proton scattering when an enhanced sample of RP_ET triggers was collected. For this particular process it was important to collect data with maximally reduced detector distance to the beamline and have continuous support of the Roman Pot experts in the control room throughout the runs. Although detector positions differ from the typical for these few runs, the active area of Roman Pots still covered the fiducial area defined for a physics measurement described in this note. We therefore preserve these runs.

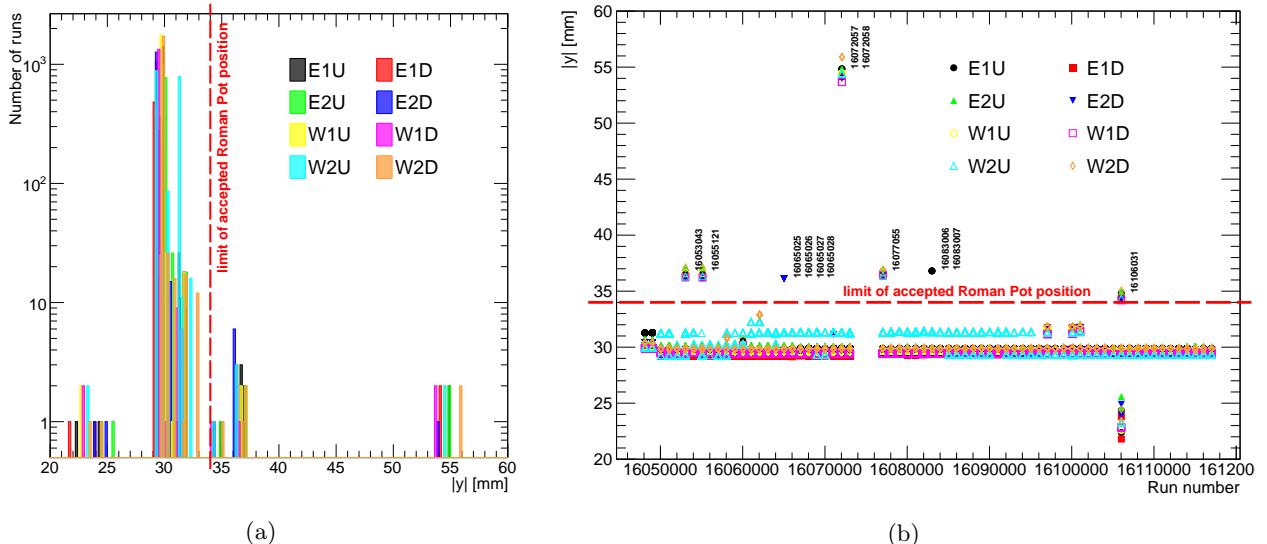


Figure 1.1: Histogram of beam-detector distance $|y|$ (1.1a) and graph showing run-dependence of $|y|$ (1.1b) for all Roman Pots.

A summary list of runs in which detector positions did not fulfill beam-detector distance limits are listed below:

16053043	16055121	16065025	16065026	16065027	16065028
16072057	16072058	16077055	16083006	16083007	16106031

Full list of runs with diffractive triggers (“RP_xxxx”) and the list of runs for each trigger can be found at the web address given in Ref. [7].

¹Roman Pots were not moved during the run. They were only moved between the runs.

2. TPC/TOF track and event selection

Charged particle tracks and primary vertices reconstructed with the TPC were selected in our analyses with the set of cuts listed below. These cuts were used in all sub-analyses described in this note and our physics analyses described in Ref. [1] and [2], unless stated differently. The goal of these criteria was to reject overlapping pile-up events, off-time pile-up tracks and ensure satisfactory resolution of the track momentum and specific energy loss.

Limit values of quantities in all selection cuts were chosen to balance the selection efficiency, background rejection power and related systematic uncertainties. Cuts in Sec. 2.2.1 and 2.2.3 were established based on histograms like the ones presented in Fig. 8.5. Cuts on z_{vtx} and track p_T and η were set based on the z_{vtx} distribution ($\sigma(z_{\text{vtx}}) \approx 50$ cm) and joint acceptance and efficiency of the TPC and TOF (see e.g. Sec. 3.2.3). More detailed description of event and track selections are contained in physics analyses notes [1, 2].

2.1 TPC vertex selection (event selection)

An event was selected for analysis if the following criteria were satisfied:

- **Single primary vertex** (exactly one that contains at least one track matched with hit in TOF),
- $|z_{\text{vtx}}| < 80$ cm.

2.2 TPC track selection

2.2.1 Quality cuts

All TPC tracks used in analyses had to satisfy quality criteria:

- $N_{\text{hits}}^{\text{fit}} \geq 25$ - at least 25 hits used in the helix fit (for good momentum resolution),
- $N_{\text{hits}}^{\text{d}E/\text{dx}} \geq 15$ - at least 15 hits used in dE/dx calculation (for good dE/dx resolution),
- $|d_0| < 1.5$ cm - transverse impact parameter w.r.t. the beamline (see Fig. 8.1) not larger than 1.5 cm (for selection of tracks of real primary particles which by definition have origin in the interaction point which lies on the beamline).

2.2.2 TOF hit matching

TPC tracks had to be matched with hits reconstructed in TOF:

- **TOF match flag** $\neq 0$.

2.2.3 Vertex matching quality

Primary TPC tracks had to match well to the primary vertex:

- $\text{DCA}(R) < 1.5$ cm - radial component of the DCA vector between the global helix and the vertex not larger than 1.5 cm (value consistent with $|d_0|$ limit),
- $|\text{DCA}(z)| < 1$ cm - absolute magnitude of longitudinal component of the DCA vector between the global helix and the vertex not larger than 1 cm.

2.3 Fiducial region of the measurement

TPC tracks had to be contained within the kinematic range:

- $|\eta| < 0.7$,
- pions: $p_T > 0.2$ GeV, tracks without identification: $p_T > 0.2$ GeV,
kaons: $p_T > 0.3$ GeV, protons: $p_T > 0.4$ GeV.

3. Efficiencies

3.1 General overview

It is best to apply the corrections to the measured distributions that are based on real data analysis. However, due to the complexity of the processes we are studying, it is necessary to obtain corrections from the MC simulation. Whenever possible, we compare efficiencies obtained from MC with these from the data and introduce appropriate corrections to the MC if necessary.

It is preferred to get the detector to true-level corrections from the MC, which is dedicated to the studied physics process. However, for this purpose, the statistics in the MC should be several times greater (preferably an order of magnitude) than we have in the data for analysis. This is not possible with generally low total efficiency (TPC & TOF & RP). On the other hand, such efficiencies depend on the modeling of the studied process and require a good description of all generally multidimensional distributions by MC. Since that, the basic method of corrections that we use in the analyses is the method of factorization of global efficiency into the product of single-particle efficiencies (separate for each central particle and forward protons). Thereby, statistically precise multidimensional corrections on TPC, TOF and RP are obtained. Nevertheless, this method does not correct on the migration effects caused by the finite detector resolution and is sensitive to the binning. Whenever possible, we use both methods of correction (at least in the consistency tests for MC samples). In this note, only the single-particle corrections are described, which are common for both analyses.

3.2 TPC track acceptance and reconstruction efficiency

We defined joint acceptance and efficiency of reconstruction of a track in the TPC, ϵ_{TPC} , as the probability that particle from the primary interaction generates signal in the detector which is reconstructed as a global track that satisfies all quality criteria (cuts 2.2.1). To derive this efficiency the single particle STARsim MC embedded into zero-bias trigger data taken simultaneously with physics triggers was used.

3.2.1 STAR nominal method

Technically, the common method used by the STAR to obtain ϵ_{TPC} is the following procedure:

1. True-level primary particles of given ID and charge were selected (*set A*).
2. Each particle from *set A* was checked if any global TPC track with more than half of hit points associated to it was generated by true-level particle (definition of true level particle-track matching, `idTruth` from `StMuTrack` collection). All particles from *set A* which have associated global track satisfying quality criteria (cut 2.2.1) formed *set B*.
3. The joint TPC acceptance and efficiency was calculated as the ratio of the histograms of true level quantities (such as p_T , η , z_{vtx}) for particles from *set B* and particles from *set A*:

$$\epsilon_{\text{TPC}}(p_T, \eta, z_{\text{vtx}}; \text{sign}, \text{PID}) = \frac{(p_T, \eta, z_{\text{vtx}}) \text{ histogram for particles of given sign and ID from } \textit{set B}}{(p_T, \eta, z_{\text{vtx}}) \text{ histogram for particles of given sign and ID from } \textit{set A}}. \quad (3.1)$$

3.2.1.1 Global track to true level particle matching

It was found during the analysis that in about 1% events there are more than one reconstructed global track matched with the same true level particle, which is shown in Fig. 3.1, separately for true level pions, kaons and protons. This generally leads to few percent overestimation of corrected number of tracks since each reconstructed track is weighted by TPC efficiency. Therefore only one reconstructed track should be considered as good track and the rest should be treated as fake tracks.

The true level particle end vertex V_r^{end} is not specified if the particle neither interacted with the dead material nor decayed. The analysis showed that 1% of the reconstructed tracks are matched to the true particle which lost identity ($V_r^{\text{end}} < 48$ cm) before entering TPC. We interpret such situation as matching of reconstructed daughter particle to true level parent particle, which is a clear bug in the STAR software rather than intentional feature. It is potentially dangerous since momentum and type of daughter particle might be significantly different from parent particle. Problem with wrong true level matching is also present for tracks with only one track matched to true level particle which do not decay or interact with material (no end vertex associated to true



Figure 3.1: Number of reconstructed global tracks, satisfying all quality criteria (cuts 2.2.1), matched with the same true level primary particle. The type of true level particle is indicated in the figure.

level particle). It is visible on Fig. 3.2 where dE/dx is shown that some reconstructed tracks have different PID than true level particle matched to it. Also, there are problems in the closure tests, where the reconstructed-level distributions of rapidity and transverse momenta weighted by the nominal efficiency corrections do not describe the true level distributions. Main reason for failing of closure tests is non-negligible (and significant for anti-protons) amount of global tracks matched with primary particles but no or little correlation in $\eta - \phi$ space between matched pair. This correlation can be represented by the distance

$$\delta^2(\eta, \phi) = (\eta^{true} - \eta^{reco})^2 + (\phi^{true} - \phi^{reco})^2 \quad (3.2)$$

between the true level particle and global track assigned to it, shown in Fig. 3.3 for particles with only one matched global track and Fig. 3.4 for particles with at least two matched global tracks, indicates that some part of the tracks taken for the efficiency calculation are measured very badly ($\delta^2(\eta, \phi)$ is very large), even if there is only one global track matched to the true-particle.



Figure 3.2: dE/dx of the track matched to true level particle. Lines indicate Bichsel function prediction for each particle species. Only tracks matched to true level particles without end vertex are shown. Black lines and arrows indicate region accepted in the analysis.

Because of several above mentioned problems with nominal STAR definition of matching between reconstructed tracks and true level particles we decided to use in the analysis modified matching definition by taking into the account the difference between reconstructed tracks and true particles in $\eta - \phi$ space.



Figure 3.3: $\delta^2(\eta, \phi)$ distributions between true level particles and tracks assigned to them. Only true level particles with only one reconstructed track matched to them were selected. Red lines and arrows indicate the cut value of 0.15^2 , which is used in the modified true level particle-track matching definition.



Figure 3.4: $\delta^2(\eta, \phi)$ distributions between true level particles and tracks assigned to them. Only true level particles with at least two reconstructed tracks matched to them were selected. Red lines and arrows indicate the cut value of 0.15^2 , which is used in the modified true level particle-track matching definition.

3.2.2 Method used in this analysis

In this method, the definition of true level particle-track matching is modified. In addition to the requirement of the appropriate number of common hit points, the distance between true level particle and track is required to be smaller than 0.15 , $\delta^2(\eta, \phi) < (0.15)^2$. It is quite an arbitrary value which should be small but not too small to loose good events. The value of δ^2 cut was chosen by the requirement that only acceptable small amount of CEP events which passed all selection criteria will not satisfy matching criteria. It was verified with

the CEP MC embedded into zero-bias triggers that with quoted value of cut on $\delta^2(\eta, \phi)$ less than 0.3% of CEP events have at least one track which is not considered to be matched with true-level pion despite the standard matching (Fig. 3.5). We consider this an acceptably low effect.

Tracks, which do not satisfy the above criterion, are treated as fake tracks (even if they are matched to the true level particle in the standard way). In 99.99% cases, where the $\delta^2(\eta, \phi) < (0.15)^2$, there is only one track matched to true level particle (Fig. 3.6). Additionally, the dE/dx of the track is mostly consistent with the true level PID (Fig. 3.7). Figure 3.8 shows the difference between TPC efficiencies obtained with the STAR standard and the modified definition of true particle-track matching. The maximum differences between TPC efficiencies in the analyzed p_T range are about 2% for pions, 3% for kaons, 2% and 4% for protons and antiprotons, respectively.

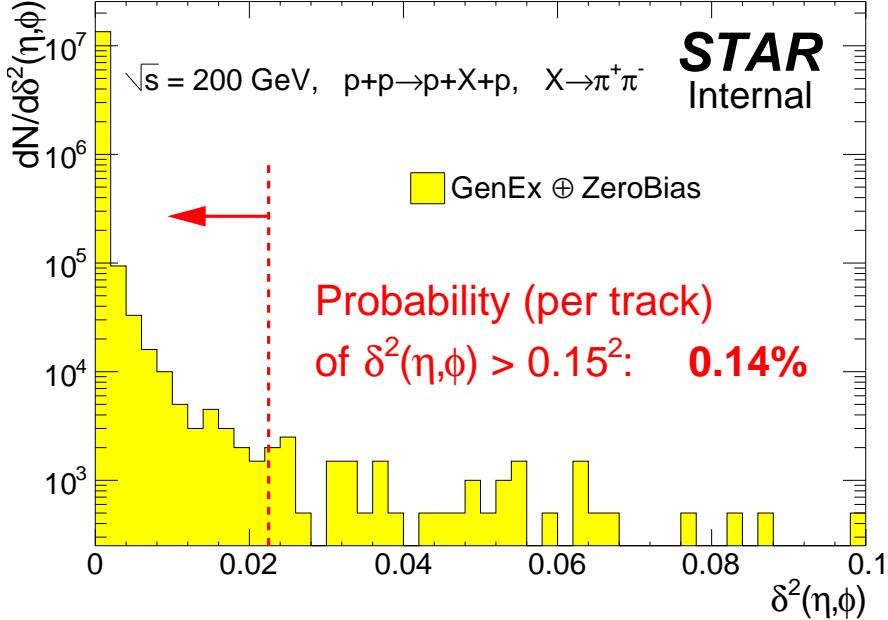


Figure 3.5: Distribution of $\delta^2(\eta, \phi)$ for tracks matched with true-level pions (using standard matching) in CEP MC embedded into zero-bias triggers. Tracks were taken from events passing full CEP event selection, recognized as exclusive $\pi^+\pi^-$. The vertical red dashed line indicates the cut value of $0.15^2 \approx 0.023$, above which less than 0.14% of tracks is contained.

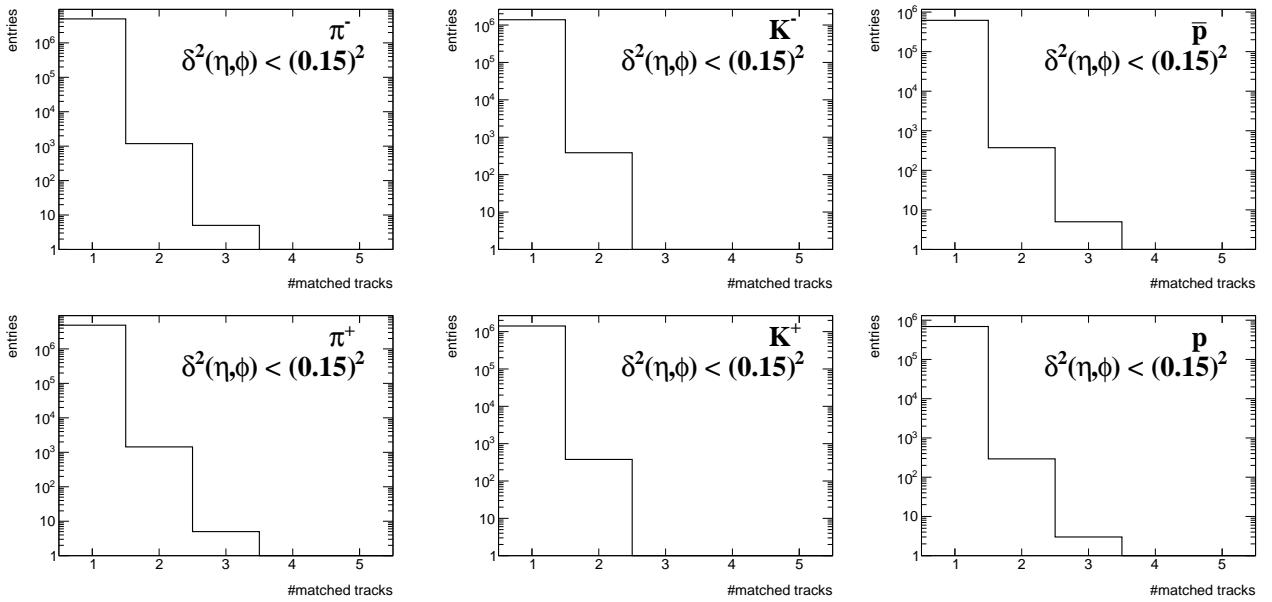


Figure 3.6: Number of reconstructed global tracks, satisfying all quality criteria (cuts 2.2.1) and $\delta^2(\eta, \phi)$ cut, matched with the same true level primary particle.



Figure 3.7: dE/dx of the closest track matched to true level particle passing the $\delta^2(\eta, \phi)$ cut. Lines indicate Bichsel function prediction for each particle species. Black lines and arrows indicate region accepted in the analysis.



Figure 3.8: TPC acceptance and reconstruction efficiency as a function of p_T ($|V_z| < 80$ cm, $|\eta| < 0.7$) obtained from two methods. Black lines and arrows indicate region accepted in the analysis.

3.2.3 Sample of efficiency plots

In Figure 3.9 we present sample plots of the TPC track acceptance and reconstruction efficiency calculated with modified definition of reconstructed track and true-level particle matching (according to description in Sec. 3.2.2), used in our analyses. Plots for all analyzed particle types and all bins of true z_{vtx} are contained in Appendix A.

In order to maximize the statistics available for the measurement (possibly wide range of accepted longitudinal vertex position z_{vtx}) with maximized probed phase-space in analyzed physics processes (wide range of track p_T and η) and minimized systematic uncertainties related to the central detector (TPC and TOF), we have studied the efficiency plots like ones shown in Fig. 3.9 and Fig. 3.10. We thus decided to set the cut on z_{vtx} at ± 80 cm, which corresponds to 89% of the full integral of normal distribution with mean at 0 and standard deviation of 50 cm. At the same time we set the cuts on track p_T and η as listed in Sec. 2.3. These cuts are represented with red dashed lines in Fig. 3.9 and Fig. 3.10. Our goal was to operate within cuboid (z_{vtx} , p_T , η) region of relatively high TPC and TOF efficiency ($\geq 50\%$ of the maximum value). In other words, we required high acceptance and efficiency for a rectangular (p_T , η) space with limits independent from z_{vtx} . One can see that the red lines in Fig. 3.9 and Fig. 3.10 always contain in their interior the region of relatively high acceptance.

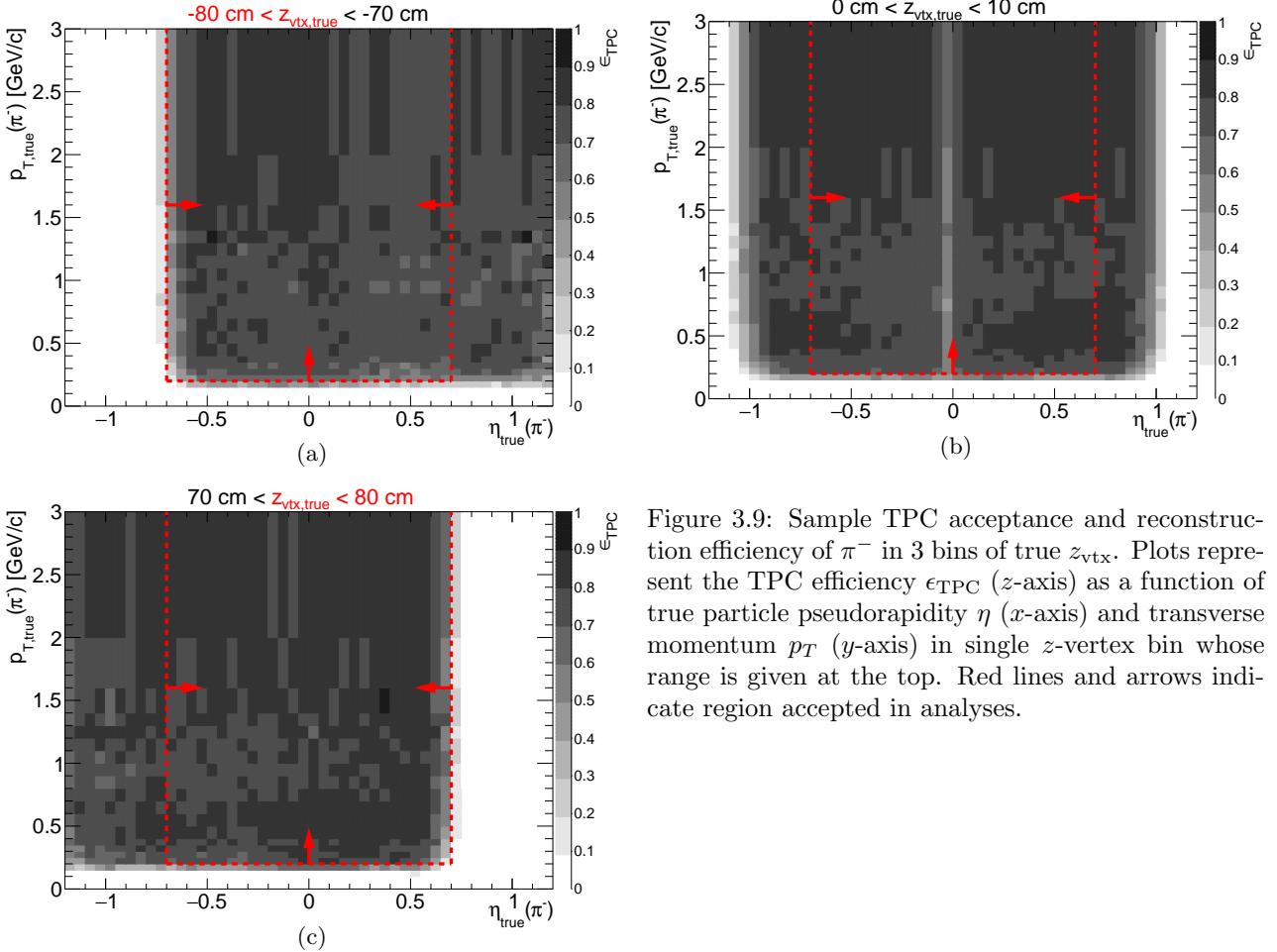


Figure 3.9: Sample TPC acceptance and reconstruction efficiency of π^- in 3 bins of true z_{vtx} . Plots represent the TPC efficiency ϵ_{TPC} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate accepted in analyses.

3.3 TOF acceptance, hit reconstruction and track matching efficiency

Combined TOF acceptance, hit reconstruction efficiency and matching efficiency with TPC tracks, ϵ_{TOF} , was defined as the probability that the global TPC track that satisfy quality criteria (cuts 2.2.1) is matched with hit in TOF (2.2.2). This quantity is generally referred as “TOF efficiency”.

It was calculated in the very similar way to TPC efficiency - single particle STARsim MC embedded into zero-bias triggers was used. Tracks belonging to *set B* from Sec. 3.2 were utilized. From these tracks a sub-sample of tracks with non-zero TOF matching flag (`StMuBTofPidTraits.mMatchFlag > 0`) was extracted

(set C). The TOF efficiency was calculated as

$$\epsilon_{\text{TOF}}(p_T, \eta, z_{vtx}; \text{sign, PID}) = \frac{(p_T, \eta, z_{vtx}; \text{sign, PID}) \text{ histogram for particles of given sign and ID from set } C}{(p_T, \eta, z_{vtx}; \text{sign, PID}) \text{ histogram for particles of given sign and ID from set } B}. \quad (3.3)$$

An additional note has to be made here about the correction which is applied to TOF matching flag in MC analysis. It was found that in embedded simulation the dead TOF elements were not masked. To correct for this effect (hence obtain more reliable TOF efficiency) a data-based map of modules was created, separately for each RHIC fill. Map was filled with modules which were matched with TPC tracks in the data. In all MC sample analyses (including efficiency determination) each TPC track with non-zero TOF match flag was additionally checked if TOF module that track was matched with had any entries in the data-based map. If not - the TOF match flag was considered 0. All dead TOF modules masked in the MC are listed in Tab. B.1 in Appendix B.

3.3.1 Sample of efficiency plots

The sample TOF efficiency plot is shown in Fig. 3.10. All remaining TOF efficiency plots are contained in Appendix B.

As shown in Sec. 4.1 the data-driven efficiency and MC efficiency differ significantly, therefore the final TOF efficiency which is used to correct the data is the one presented here (taken from single particle embedded MC) but additionally modified according to correction derived in the referred section.

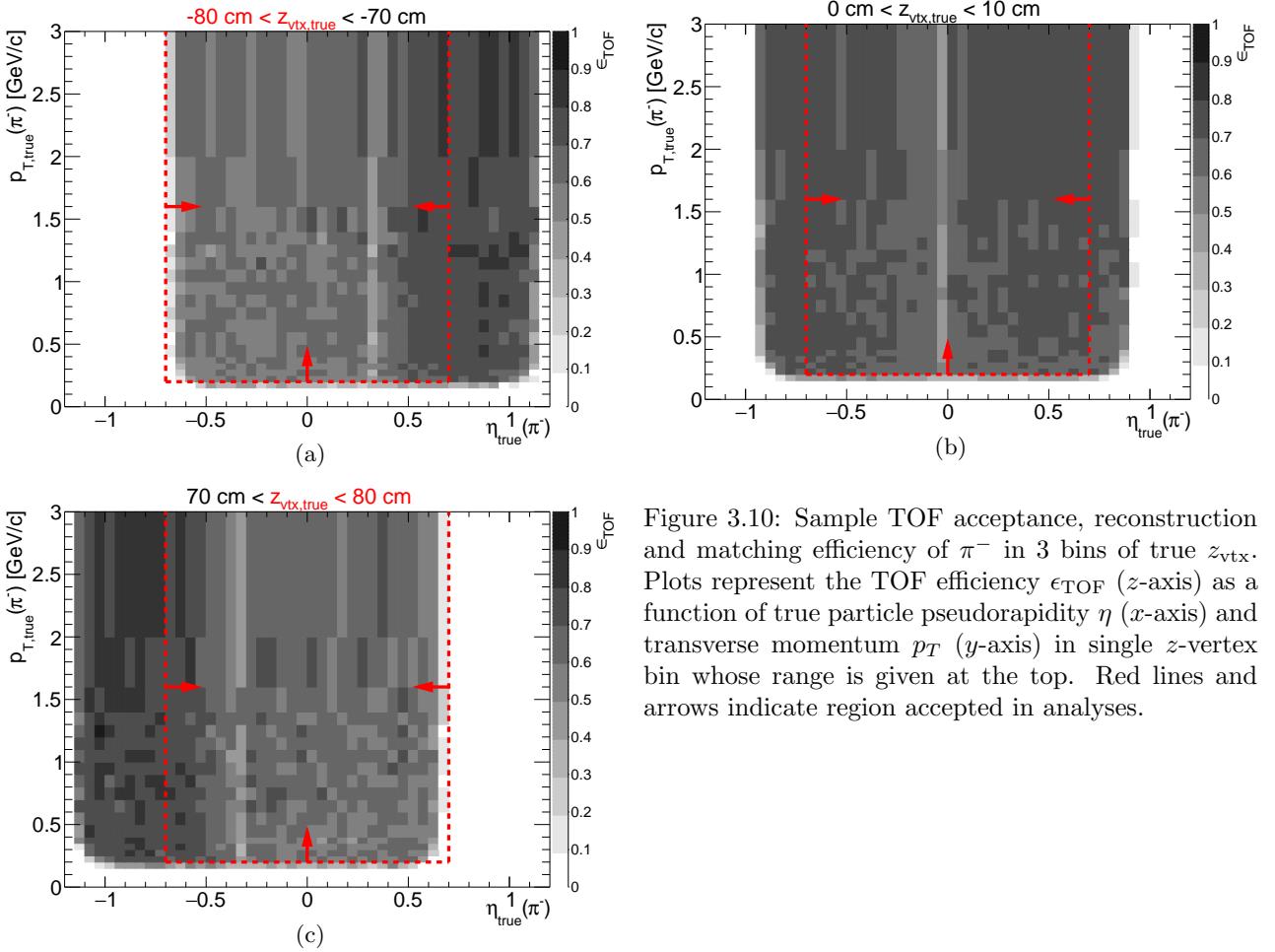


Figure 3.10: Sample TOF acceptance, reconstruction and matching efficiency of π^- in 3 bins of true z_{vtx} . Plots represent the TOF efficiency ϵ_{TOF} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.

4. Corrections to TPC/TOF efficiency

4.1 Data driven corrections to TOF efficiency

The efficiency of TOF hit reconstruction and matching with the TPC tracks that was used in our analyses was (at the very beginning) taken directly from the STAR simulation. This made the inaccuracies in the description of real detector geometry and its response propagating to physics results and introducing a bias. We decided to derive the correction to the TOF efficiency obtained from the STAR simulation by extracting it in the very same way from the data and embedded MC and comparing the results.

Unfortunately, in 2015 there were no low-luminosity (heavy-ion) runs that would imply lack of off-time pile-up tracks in TPC and thus would allow calculating the TOF efficiency in straightforward way, namely by dividing number of selected TPC tracks that were matched with TOF hits by number of all selected TPC tracks (matched or unmatched). For this reason a variation of the “tag and probe” method was developed and used. This method utilizes some specific feature of the distribution of quantity describing two objects (whose trigger/reconstruction/identification/etc. efficiency is studied) which allows to quantify amount of these objects with satisfied/unsatisfied efficiency condition. This could be e.g. J/ψ peak in the invariant mass spectrum of the muon tracks, like in the study of muon identification efficiency in MTD [8].

In our variation of mentioned tag&probe method the CEP of $\pi^+\pi^-$ events were used, with the missing transverse momentum p_T^{miss} used to determine signal event yield ($p_T^{\text{miss}} = (\vec{p}_W + \vec{\pi}^+ + \vec{\pi}^- + \vec{p}_E)_T$). In short, events with forward proton track on each side of STAR and with a TOF-matched primary TPC track (tag) were selected. Among the remaining primary TPC tracks in the same vertex the opposite-sign TPC track (probe) was chosen as the one which provides the minimum total transverse momentum of all four tracks. The probe was checked whether it has been matched with the TOF hit or not. The ratio of the matched TPC tracks to all (matched or unmatched) TPC tracks defined the TOF efficiency, as given in Eq. (4.1):

$$\varepsilon^{\text{TOF}} = \frac{N_{\text{satisfied}}^{\text{probes}}}{N_{\text{produced}}^{\text{probes}}} = \frac{N_{\text{satisfied}}^{\text{probes}}}{N_{\text{satisfied}}^{\text{probes}} + N_{\text{failed}}^{\text{probes}}} = \frac{2N_{TT}^{\text{events}} + N_{TP_s}^{\text{events}}}{2N_{TT}^{\text{events}} + N_{TP_s}^{\text{events}} + N_{TP_f}^{\text{events}}} \quad (4.1)$$

Subscripts TT , TP_s and TP_f denote possible combinations of tags and probes:

- TT (both tracks satisfy tag criteria, and by definition the efficiency condition); such events provide two probes which satisfy efficiency condition which is the reason for the factor ‘2’ in front of N_{TT}^{events} in Eq. (4.1),
- TP_s (one track is tag, the other is a probe and probe satisfies the efficiency condition),
- TP_f (one track is tag, the other is a probe and probe fails to satisfy the eff. condition).

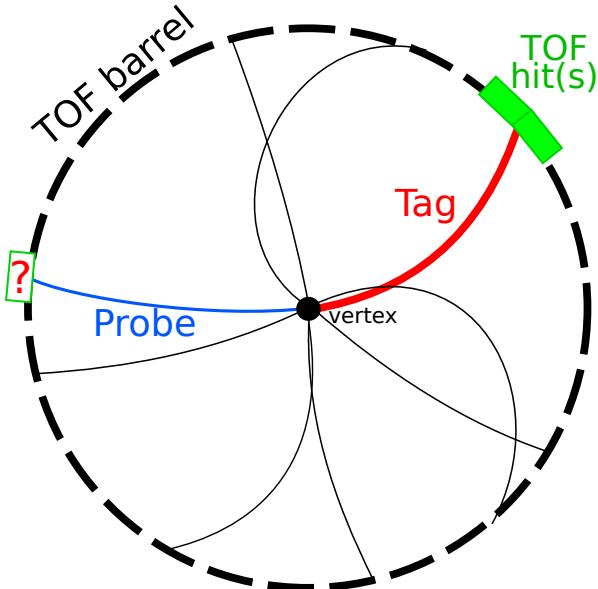


Figure 4.1: Sketch of the cross section of the central detector and CEP event with off-time pile-up tracks with drafted tag&probe method used to determine the TOF hit reconstruction and matching efficiency.

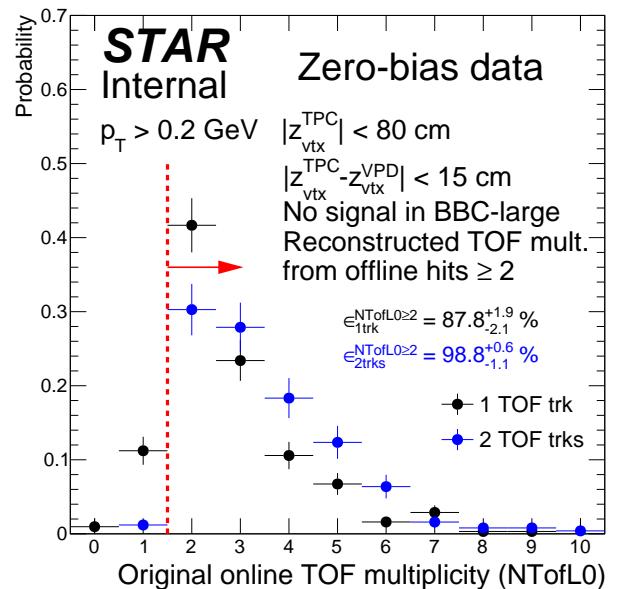


Figure 4.2: TOF L0 multiplicity in events with exactly 1 (black) or exactly 2 (blue) primary TPC track matched with offline TOF hits under condition of online TOF mult. reconstructed from offline hits ≥ 2 .

The method is illustrated in the sketch in Fig. 4.1. The detailed description of the procedure of the TOF efficiency extraction is listed below:

1. Data from RP_CPT2 trigger were used. This trigger required at least 2 level-0 TOF hits (trigger details can be found in Tab. 2.1 of Ref. [1]). Requirement of 2 TOF hits in the trigger (to ensure presence of 2 tracks in TPC) does not comply with the basics of the tag&probe method, but as explained in #4 and in the following page the tight selection of tags as tracks providing online multiplicity 2 and firing the TOF trigger allowed to use this method. GenEx $\pi^+\pi^-$ signal MC (embedded into zero-bias data from runs corresponding to RP_CPT2 triggers), as well as Pythia CD for non-exclusive background, were subjected to the same trigger conditions, except online TOF multiplicity condition to avoid trigger bias.
2. Events were selected with the cuts from nominal CEP analysis (Ref. [1]), except cuts **C3** and **C7-C9** which were removed or modified as given below. This provided significant reduction of background events.
3. For each event passing above selection primary TOF-matched TPC tracks of good quality (cuts 2.2.1,2.2.3), contained within kinematic region of our measurement (cuts 2.3 with p_T threshold lowered to 0.18 GeV to also calculate efficiency in a point where it rapidly changes hence allow more precise fit to efficiency p_T -dependence) and compatible with pion hypothesis based on dE/dx ($|n_{\text{pion}}^\sigma| < 3$) were selected. If any TOF-matched track incompatible with pion hypothesis was found, event was dropped from analysis. Also, event was not analyzed if more than 2 TOF-matched tracks were reconstructed (not two-prong CEP event).
4. Primary TOF-matched TPC tracks preselected in #3 were set as tag candidates. In the data they were additionally required to be matched with the TOF hit belonging to a TOF cluster that is expected to provide multiplicity 2 at the level 0 (to reduce trigger bias). In case only one track was found to be a tag candidate, steps #5-#6 were executed only with this single track being a tag. Otherwise steps #5-#6 were repeated for each of two tracks set as a tag.
5. From the remaining TPC tracks in the same vertex (preselected in #3) of the sign opposite to tag and of good quality (cuts 2.2.1,2.2.3), the one which provided the best transverse momentum balance together with 2 protons and a tag was selected as a probe (signature of exclusive $\pi^+\pi^-$ is $p_T^{\text{miss}} \sim 0$). If no primary TPC tracks passing this selection were found, an event was dropped.
6. 2-dimensional histograms of quantities of interest q (probe η , probe p_T , z_{vtx}) vs. p_T^{miss} were filled, separately for all probes and only for probes matched with TOF. Each probe (each entry to the histogram) was associated with the weight w taking into account the trigger efficiency and vertexing efficiency, as given in Eq. (4.3) and explained later in the text.
7. In each bin of quantity of interest q (as a function of which the efficiency was to be determined) the distribution of p_T^{miss} was fitted in the signal-free region with the function describing background shape. The background function was extrapolated to $p_T^{\text{miss}} = 0$. The signal yield in given bin of q was calculated as the integral of the histogram with subtracted integral of the background function, both in the range $p_T^{\text{miss}} < 75$ MeV. The final efficiency in given bin of q was calculated according to Eq. (4.2):

$$\varepsilon_{\text{TOF}}(q) = \frac{N_{\text{weighted}}^{\text{matched}, p_T^{\text{miss}} < 75 \text{ MeV}} - N_{\text{bkgd,weighted}}^{\text{matched}, p_T^{\text{miss}} < 75 \text{ MeV}}}{N_{\text{weighted}}^{\text{all}, p_T^{\text{miss}} < 75 \text{ MeV}} - N_{\text{bkgd,weighted}}^{\text{all}, p_T^{\text{miss}} < 75 \text{ MeV}}} \quad (4.2)$$



Figure 4.3: Sample distributions of total transverse momentum p_T^{miss} of the $p+\text{Tag}+\text{Probe}+p$ system in the data (4.3a) and signal+background embedded MC (4.3b). TOF-matched and all (matched or unmatched) probes are represented by histograms in green and black color, respectively. The red dashed line shows the exclusivity cut value (75 MeV/c). Signal yield is determined via the integral of the histogram with subtracted integral of the solid line representing non-exclusive background in the p_T^{miss} range to the left from the vertical line. Background (solid line) is fitted with 2nd order polynomial in the signal-free region (the details can be found in Ref. [1] in Sec. 5.2). Full sets of histograms in different ranges of probe p_T and η are included in Appendix G.

As described in the algorithm of the tag&probe method, the RP_CPT2 triggers were used in our study. The logic of that trigger required at least 2 TOF hits online. Since the system whose efficiency was studied has also been a part of the trigger, the tag should be, in principle, chosen as the track that is linked with 2 online TOF hits - to be sure that the tag satisfies the trigger condition and thus the probe is not biased by the trigger (in other words, the probe does not bias the resultant efficiency). Unfortunately, the TOF system works independently for the trigger and for the offline data stream (the readout electronics are independent), therefore there is no information about the connection between the TOF hits at L0 and offline.

Fortunately, a solution to the above limitation has been found. We made use of the TOF cluster¹ concept introduced in Ref. [1], Sec.XX, as well as of reconstruction/emulation of the online TOF multiplicity from the offline hits according to description of the TOF trigger system in Ref. [9]. With these tools we chose tags as the tracks that are matched with TOF hit associated with a TOF cluster that provides reconstructed online multiplicity ≥ 2 . We verified with zero-bias data that by requiring the reconstructed online multiplicity to be ≥ 2 we can efficiently select events in which the true online (L0) TOF multiplicity was also ≥ 2 . For this purpose we selected events with only one(two) TOF-matched TPC track(s), in addition forming a primary vertex. We also required no signal in large BBC tiles, the same as in physics analysis of CEP. To make sure that the single TOF-matched track is from beam-beam interaction that TOF electronics is adjusted to trigger on, we required the signal in VPD detectors on both sides of the IP. The z -positions of the primary vertex reconstructed in TPC and reconstructed from the time difference in west and east VPDs was required to be not larger than 15 cm. In Fig. 4.2 we show distribution of the TOF L0 multiplicity for such selected events. One can find in this figure that the probability of TOF L0 multiplicity ≥ 2 for a single TOF-matched track, $\epsilon_{1\text{trk}}^{\text{NTofL0} \geq 2}$, is high and equals 87.8%. For events with 2 reconstructed TOF-matched tracks the probability of TOF L0 multiplicity ≥ 2 , $\epsilon_{2\text{trks}}^{\text{NTofL0} \geq 2}$, equals 98.8%. These efficiency factors were accounted in the event weighting, as shown in Eq. (4.3).

The vertexing efficiency was an additional efficiency factor which had to be taken into account in described analysis². This efficiency for a single track depends on matching with TOF due to 'useBTOFmatchOnly' option used in reconstruction. It is different when both exclusive pion tracks are matched with TOF, comparing with the case when only one track is matched with TOF and the other is not. The vertexing efficiency defined as the probability that two global TPC tracks of true-level primary particles, matched with TOF and satisfying criteria 2.2.1, form the common primary vertex is presented in Ref. [1] in Fig.XXX as a function of the distance in z of the DCA points on the beamline, $|\delta z_0|$. This efficiency extracted from embedded MC was used in this analysis in case when both tag and probe were matched with TOF. As shown in dedicated study presented in



Figure 4.4: Efficiency $\epsilon_{\text{vtx}}^{\text{nTOF}=1}$ of reconstruction of the primary vertex from a single TOF-matched TPC track as a function of the absolute value of the transverse impact parameter d_0 (4.4a) and efficiency $\epsilon_{\text{vtx}}^{\text{no-TOF}}$ of association with the vertex formed from single TOF-matched track of a TPC track not matched with TOF, as a function of radial and longitudinal DCA to the vertex (4.4b). The tracks were required to be matched with true-level primary particles (pions) as well as the quality and kinematic cuts were implied (2.2.1, 2.2.3, 2.3).

¹Group of offline TOF hits adjacent/close to each other in space (and time)

²This analysis could be performed using solely global TPC tracks without requiring reconstructed primary vertex. Such solution was not chosen due to limitation of the picoDST data accessible for analysis (not all global tracks stored).

Ref [1] in Sec.XXX, the MC reproduces very well the vertexing efficiency obtained from the data.

For events with only one of exclusively produced pions being matched with TOF a different efficiency had to be used. In this case only the TOF-matched track participated in the vertexing, while the other was added to the primary vertex if it satisfied certain criteria. We therefore calculated the vertexing efficiency for the single TOF-matched track as a function of the transverse distance to the beamline $|d_0|$ (Fig. 4.4a) and the efficiency of association of the TPC track not matched with TOF with the primary vertex made of single TOF-matched track (Fig. 4.4b). The final correction factor for a probe represented by the weight of an entry to q vs. p_T^{miss} histograms has the following form:

$$w = \begin{cases} \left[\epsilon_{1\text{trk}}^{\text{NTofL0} \geq 2} \times \epsilon_{vtx}^{\text{nTOF}=2}(|\Delta z_0|) \right]^{-1} & \text{for } TT, TP_s \\ \left[\epsilon_{2\text{trks}}^{\text{NTofL0} \geq 2} \times \epsilon_{vtx}^{\text{nTOF}=1}(|d_0|) \times \epsilon_{vtx}^{\text{no-TOF}}(\text{DCA}_z, \text{DCA}_R) \right]^{-1} & \text{for } TP_f \end{cases}, \quad (4.3)$$

where factors related to trigger efficiency are accounted only in the data analysis. A sample histograms of the p_T^{miss} distribution in the data and embedded MC filled with entries of weight w are shown in Fig. 4.3.

The dependence of the TOF efficiency on the track p_T was assumed to have the error function (erf) form:

$$\epsilon^{\text{TOF}}(p_T) = \frac{1}{2} \cdot P_1 \cdot \left(\text{erf}\left(\frac{p_T - P_2}{P_3}\right) + 1 \right) \quad (4.4)$$

In Fig. 4.5 we show the resultant TOF efficiency obtained with the Tag&Probe method.

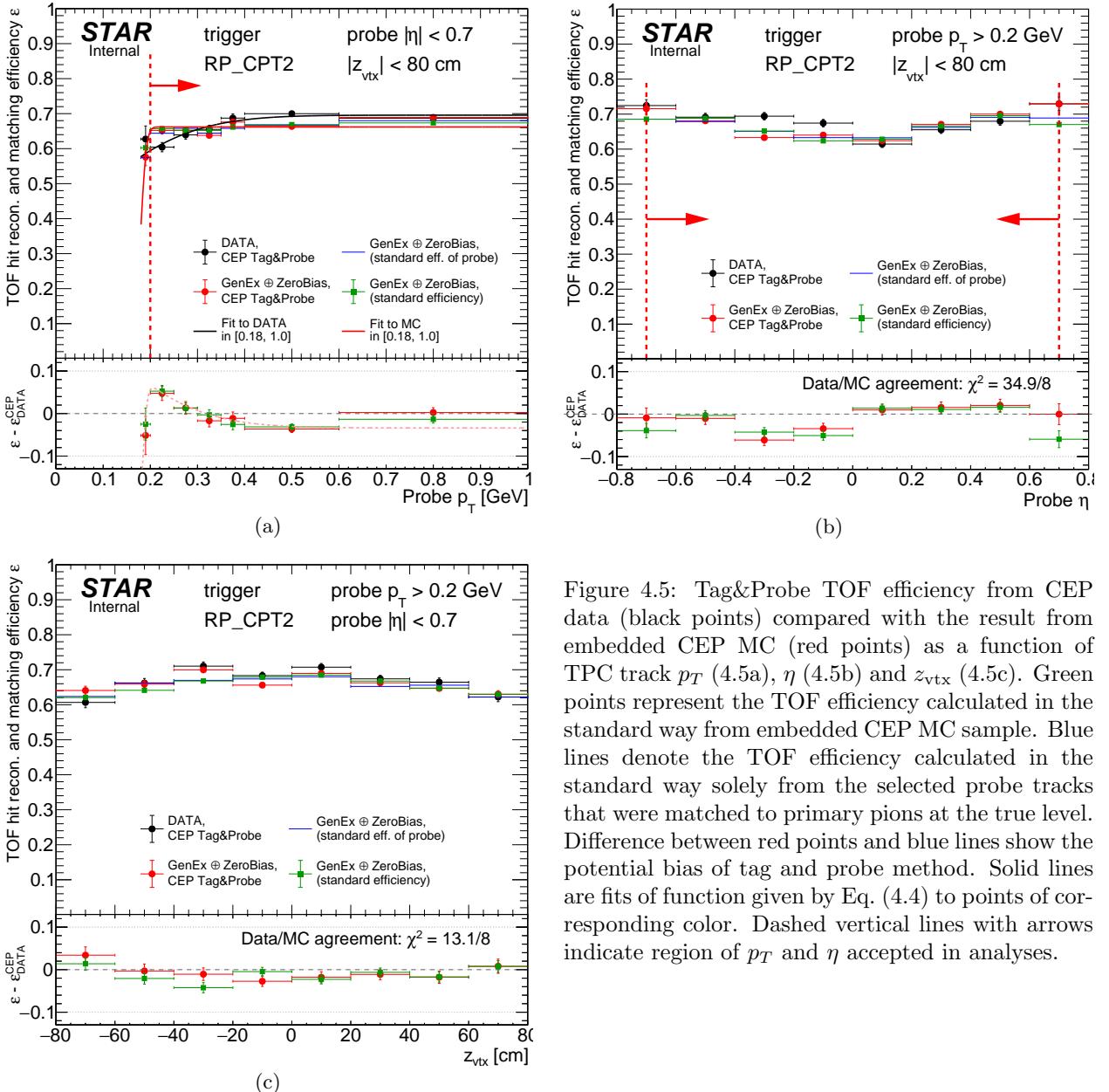


Figure 4.5: Tag&Probe TOF efficiency from CEP data (black points) compared with the result from embedded CEP MC (red points) as a function of TPC track p_T (4.5a), η (4.5b) and z_{vtx} (4.5c). Green points represent the TOF efficiency calculated in the standard way from embedded CEP MC sample. Blue lines denote the TOF efficiency calculated in the standard way solely from the selected probe tracks that were matched to primary pions at the true level. Difference between red points and blue lines show the potential bias of tag and probe method. Solid lines are fits of function given by Eq. (4.4) to points of corresponding color. Dashed vertical lines with arrows indicate region of p_T and η accepted in analyses.

From Fig. 4.5 one can find that the efficiency in the data and MC differ substantially. The comparison of the p_T -dependence of TOF efficiency in Fig. 4.5a suggests that saturation of the TOF efficiency with growing track p_T is slower than MC predicts, and that maximum (high- p_T) efficiency is generally higher by 3-4 percentage points in the data compared to MC. In the η -dependence we can see that there are some bins of η in which the agreement between MC and data is satisfactory, while in others the difference is large. The z_{vtx} -dependence (Fig. 4.5c) is in acceptable agreement between the data and MC.

The above observations led to perform the analysis of p_T -dependence of TOF efficiency in 4 bins of track pseudorapidity. The result is presented in Fig. 4.6. Parameters of functions fitted to data and MC points are contained in Tab. 4.1. The additive correction to the TOF efficiency presented in Sec. 3.3 has the form

$$\delta\varepsilon^{\text{TOF}}(p_T) = \varepsilon_{\text{DATA}}^{\text{TOF}}(p_T) - \varepsilon_{\text{MC}}^{\text{TOF}}(p_T), \quad (4.5)$$



Figure 4.6: Tag&Probe TOF efficiency from CEP data (black points) compared with the result from embedded CEP MC (red points) as a function of TPC track p_T for four ranges of the pseudorapidity η of the probe. Blue lines denote the TOF efficiency calculated in the standard way solely from the selected probe tracks that were matched to primary pions at the true level. Difference between red points and blue lines show the potential bias of tag and probe method. Solid lines are fits of function given by Eq. (4.4) to points of corresponding color, whose parameters are tabulated in Tab. 4.1. Dashed vertical lines with arrows indicate region of p_T and η accepted in analyses.

η range	Data			MC		
	P_1	P_2	P_3	P_1	P_2	P_3
$[-0.7; -0.3]$	0.736 ± 0.011	0.013 ± 0.083	0.329 ± 0.106	0.686 ± 0.010	0.180 ± 0.036	0.011 ± 0.039
$[-0.3; 0.0]$	0.696 ± 0.011	0.090 ± 0.070	0.164 ± 0.079	0.629 ± 0.003	0.180 ± 0.004	0.029 ± 0.011
$[0.0; 0.3]$	0.716 ± 0.031	-0.050 ± 0.137	0.508 ± 0.233	0.670 ± 0.005	-0.021 ± 0.096	0.344 ± 0.154
$[0.3; 0.7]$	0.683 ± 0.008	0.132 ± 0.055	0.084 ± 0.056	0.690 ± 0.009	0.167 ± 0.002	0.000 ± 0.000

Table 4.1: Parameters of function (Eq. (4.4)) describing the TOF efficiency as a function of pion track p_T in bins of track η from Fig. 4.6. Paramaters of the functions fitted to both data and MC points are given. The TOF efficiency calculated as described in Sec. 3.3 is corrected during data analysis by the difference of values of functions $\varepsilon^{\text{TOF}}(p_T)$ with parameters for data and MC, respectively (the difference is added).

where $\varepsilon_{\text{DATA}}^{\text{TOF}}(p_T)$ and $\varepsilon_{\text{MC}}^{\text{TOF}}(p_T)$ are of the form given by Eq. (4.4) with parameters from Tab. 4.1 for the data and MC, respectively. The same correction is used for positive and negative charge particles.

Due to much lower statistics a similiar study cannot be performed for kaons and protons. We therefore apply the same correction $\delta\varepsilon^{\text{TOF}}(p_T)$ to the TOF efficiency for K^\pm and $p(\bar{p})$.

5. Energy Loss Correction

Particles passing through the detector material lose energy as they travel. The track momentum p_T is reconstructed by fitting a helical path to the track points left in the detector. Fitting the track points to an ideal helical track tends to underestimate the momentum due to these energy loss effects. To minimize biases due to this effect, correction procedure is applied during standard track momentum reconstruction for both data and MC simulation. For this procedure pion hypothesis is used and the reconstructed momentum p_T^{meas} is corrected by the amount of energy loss specific for a pion. For all particles but a pion some rest bias is still present since energy loss is specific for each particle type. These biases can be determined from simulated tracks run through GEANT. The correction $p_T^{meas} - p_T^{true}$ was calculated for each particle species as a function of p_T^{meas} , η and z -vertex. The sample energy loss correction averaged over $|\eta| < 0.7$ for K^- is shown in Fig. 5.1. The energy loss corrections for other particle species are shown in Figs. C.1 to C.10 in Appendix C.

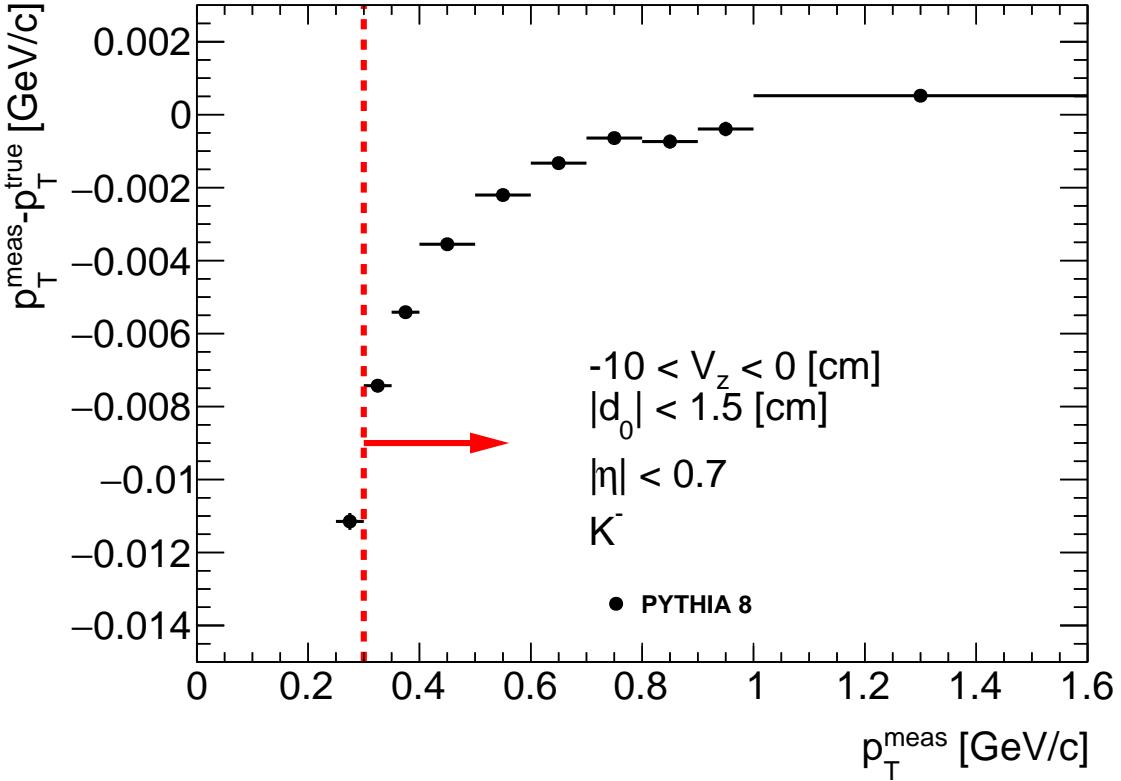


Figure 5.1: Sample energy loss correction $p_T^{meas} - p_T^{true}$ for K^- as a function of reconstructed transverse momentum p_T^{meas} ($|\eta| < 0.7$) in single z -vertex bin, $-10 < V_z < 0$ cm. Red lines and arrows indicate region accepted in analyses.

6. Forward protons

One of the main detector systems used in our analyses is the Roman Pot setup in Phase II* configuration (Roman Pots in Phase I took data with STAR during special runs in last days of $p+p$ collisions in 2009, see e.g. Ref. [10–12]). It allowed to trigger on forward protons and reconstruct their momentum with high efficiency and precision.

6.1 Detector layout

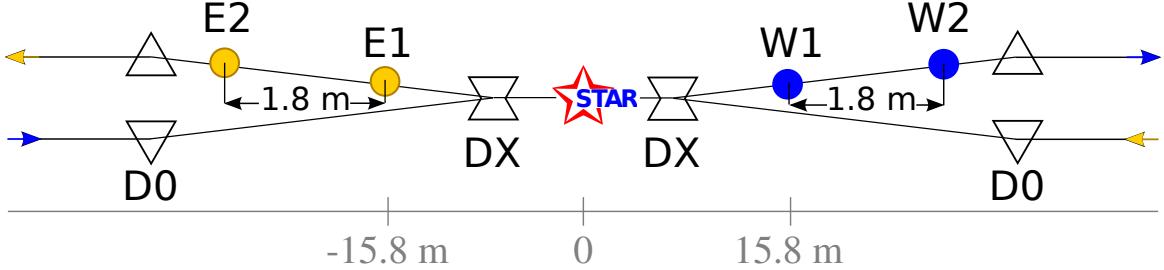


Figure 6.1: Schematic representation (top view) of the Roman Pot Phase II* at STAR (not to scale).

As presented in Figure 6.1 the Roman Pot Phase II* setup consists of detectors located in two stations on each side of the interaction point (IP) in a distance of 15.8 m and 17.6 m from the IP. Each station has two Roman Pots positioned vertically, one above and the other below the beamline (Fig. 6.2). Detectors are situated downstream the DX dipole magnets responsible for head-on targeting of the incoming beams and bending outgoing beams back into the accelerator pipeline. The constant and uniform magnetic field of the DX magnet works as a spectrometer and thus knowledge on the track angle and position in the detector allows complete reconstruction of the proton momentum, including the fractional momentum loss ξ (as described in [13]). The naming convention of elements of RP setup is described in Ref. [14].

Single Roman Pot (the vessel, Fig. 6.5) houses a package of 4 silicon strip detector planes (Fig. 6.6) - one pair of SSDs with vertical and one with horizontal orientation of the strips, and hence measurement of the position of a proton hit is possible in both transverse spatial coordinates, x and y . The pitch (distance between neighbouring strips) in a single detector is $100 \mu\text{m}$, therefore intrinsic spatial resolution is at the level of $100/\sqrt{12} \approx 30 \mu\text{m}$. In addition to the silicon detectors, the package contains plastic scintillator that covers whole active area of the silicon, attached at the back. Two lightguides are glued at the top edge of scintillator which direct the light generated when ionizing particle passes through it to the photomultiplier tubes (PMTs) connected at the very end of each. This counter is used to trigger on forward protons and also provides the timing information.

6.2 Roman Pot data reconstruction

6.2.1 Roman Pot track points and tracks

Roman Pot data is stored in MuDST in StMuRpsCollection. This class contains objects reconstructed with St_pp2pp_Maker. The basic (low-level) data objects are the clusters characterized by their length (number of adjacent strips with signal above the threshold), energy (sum of ADC counts in each constituent strip) and

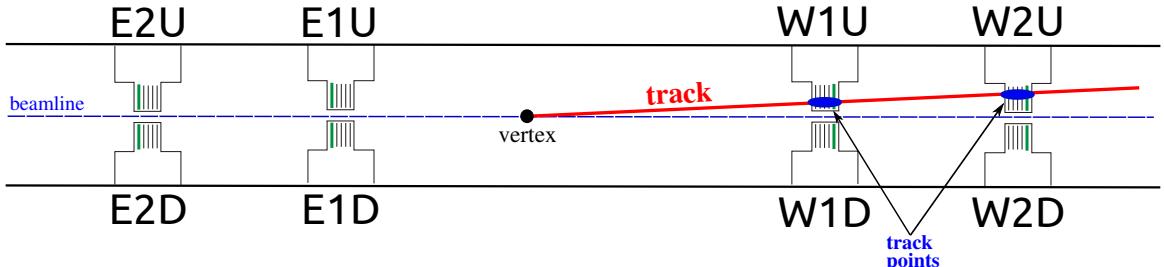


Figure 6.2: Side view of the Roman Pot Phase II* setup (not to scale) with an illustration of Roman Pot track point and track.

position. Vectors of clusters are provided independently for each silicon plane. Another low-level data are informations about time (TAC) and signal strength (ADC) for each PMT.

Our physics analyses utilized mainly the high-level objects which are the track points (StMuRpsTrackPoint) and tracks (StMuRpsTrack) stored in vector members of StMuRpsCollection. In short, these objects represent real particles (e.g. their momentum vector) in the same way as the TPC tracks represent particles traversing TPC. Concept of track point and track is depicted in Fig. 6.2 and described in some more details in Ref. [15].

The algorithm for RP track reconstruction is implemented in St_pp2pp_Maker. It is a multi-track algorithm which first forms track points from clusters (there may be many track points in single RP), and then form tracks from all possible combinations of track points in Roman Pots in the same branch [15,16]. The track points and tracks are additionally tuned with Roman Pot “afterburner” package (StMuRpsUtil, [17]) which recalculates positions of hits and momenta of tracks according to the final alignment corrections and known vertex position.

6.2.2 Alignment

Precise knowledge of positioning of detectors in space is crucial for correct reconstruction of proton momentum. Therefore a process of detector alignment was done, which involved a few steps. At first, dedicated detector survey [4] was performed before the start of run 15, which provided initial calibration of the LVDTs installed



Figure 6.3: Correlation between the hit position of constituent track point in the first RP station (y -axis) and the local angle of track (x -axis) in elastic scattering events. The y -intercept has interpretation of the average position of the interaction vertex in given coordinate.

in Roman Pot movement system. This was sufficient to know the positioning of detectors at a $\gtrsim 1$ mm level. Next, the alignment analysis using elastic scattering events was done, as described in the analysis note of elastic proton-proton scattering [18]. This analysis provided detectors alignment in the so called “elastic scattering reference frame”, by design close to the global (STAR) reference frame. In the last step determination of the average vertex position was done, as described below.

Vertex position is not necessary to correctly reconstruct forward proton observables in elastic scattering events e.g. squared four-momentum transfer $|t|$ - this is because one can use momentum balance constraint of elastically scattered protons (collinearity constraint) and calculate scattering angle from the straight line fit to all track points of east and west proton tracks, without knowledge where the interaction vertex was. The same approach cannot be used in other processes like Single or Central Diffraction, since there is only single forward proton (SD) or two forward protons (CD, CEP) are independent in terms of scattering angles and momentum loss.

Need for a vertex position in the reconstruction of non-beam-momentum proton tracks led to development of the method of extraction of the average vertex position using the elastic scattering data, as presented in Ref. [19]. For this purpose RP_ET triggers from randomly chosen runs (16085056 and 16085057) were used with single (exactly one) global RP track reconstructed on each side of the IP. Tracks were required to be collinear at 2σ level. Selected clean sample of elastic scattering events was used to prepare the plots of the position of the track point in near RP station vs. the local angle of the RP track with respect to the global z -axis (Fig. 6.3). The least squares fit of a line (with perpendicular offsets) to all data points in the scatterplot was performed. As a result four lines were obtained, one per arm per transverse coordinate. The slope of the line has interpretation of the distance from the nominal IP ($z = 0$) to the 1st RP station at 15.8 m. One can see that the slopes are well consistent with this value. The intercept of the line equals to the average position of the vertex in given coordinate. One finds that $\langle x \rangle_{IP}$ obtained from the fits to data points in two independent elastic arms are perfectly consistent, while in $\langle y \rangle_{IP}$ parameters differ by 1.5 standard deviations. We conclude that extracted values of average positions of the vertex are trustworthy and we can average the numbers obtained from two independent arms. As a result we use in our analyses numbers $\langle x \rangle_{IP} = 0.42$ mm, $\langle y \rangle_{IP} = 0.45$ mm, both in the reconstruction/recalculation of proton tracks with StMuRpsUtil package, and generation of MC events.

In Ref. [19] the method of $\langle z \rangle_{IP}$ extraction is also presented, however the result $\langle z \rangle_{IP} = 3$ cm is much smaller than the distance between nominal IP and the Roman Pots (3 cm compared to 15.8 m). Such small offset does not bias proton momentum reconstruction, therefore we neglect it.

Due to time dependence of the beam conditions, automatic beam orbit corrections etc., the average position of the vertex may change from run to run. The measure of this variation is a by-product of Roman Pot alignment procedure which was done for every run. In Ref. [20] the middle points of the track (MPTs, position of the track at $z = 0$) are plotted as a function of run number. One can see that MPTs in x and y are roughly constant along the entire data taking period in 2015 and consistent with numbers derived in Fig. 6.3. Another cross-check for the correctness of extracted $\langle x \rangle_{IP}$ and $\langle y \rangle_{IP}$ was done with the use of elastic scattering MC simulation in Geant4. Several MC samples were generated with differently positioned vertex. The same procedure was performed as the one described in this section and the output values of $\langle x \rangle_{IP}$ and $\langle y \rangle_{IP}$ were always consistent with the true values. Also the comparisons of the hit maps of elastic scattering protons were done between the data and MC, and the best matching was found for vertex generated at $\langle x \rangle_{IP} = 0.42$ mm and $\langle y \rangle_{IP} = 0.45$ mm (e.g. [21]).

6.3 Roman Pot simulation

6.3.1 General description

Simulation of the STAR detector (STARsim) implemented in Geant3 does not contain a model of the Roman Pot detectors. Because of this, a dedicated simulation program ”pp2pp” was prepared to enable precise measurements with the Roman Pot data. The development of this software started in 2012 as two independent projects for “Geant4 simulation tool kit” subject at AGH UST, which were later included in B.Sc. theses of the authors. One project was devoted to modeling of the Roman Pot and Silicon Strip Detector package [22], the other was aimed to implement simulation of the collider elements with full magnet lattice (including B field) [23] between the IP and Roman Pot location in Phase I configuration (at $z = \pm 55$ m) based on the MAD-X twiss files. At that stage the two programs were used in analyses of 2009 $p + p$ data [12, 24] taken with Roman Pots during special high- β^* runs. Both projects were merged in 2013 and since then are continuously developed by the Krakow group.

Currently the simulation program is a standalone Geant4 application that allows to simulate any type of particle, track it from the IP to the Roman Pots location (either in Phase I/run 2009 or Phase II*/run 2015&17 configuration) and obtain the true level and reconstructed information in the standard STAR data format - MuDST classes stored the ROOT tree. In the reconstruction the original STAR St_pp2pp_Maker class is used. The program has multiple generation options: the input can be directly from STARsim, Pythia, HepMC format or a simple text file. There are also built-in modes, such as elastic scattering event generation. One can

customize the beam conditions, magnet settings (beam energy), Roman Pot positions, and many others. The application is installed at RCF and free to use by any STAR collaborator. Implementation of extensions desired by analyzers is possible after contact with authors. Some more information about the Roman Pot simulation (geometry, usage) can be found in the presentation from the LFS-UPC PWG weekly meeting (Ref. [25]).

6.3.2 Detector model

The geometry of the collider elements (beampipe, magnets) and Roman Pot detectors (vessel, SSD packages) is implemented in Geant4 according to the best available sources of information. The dimensions, positioning and material of the beam pipe and magnets were taken from the technical drawings available at RHIC C-AD and were discussed with the designers of these elements. In case of Roman Pot vessels and detector packages not only technical drawings were used to find the appropriate dimensions, but also dedicated survey was made in which some elements were measured with caliper.



Figure 6.4: View of the Geant4 implementation of the Roman Pot Phase II* detector setup.

In Fig. 6.4 we show the general view of the "world" volume in Geant4 pp2pp simulation with labeled the most significant elements of the geometry. In Fig. 6.7 we show a few shots of the Roman Pot housing and the SSD package implemented in Geant4, together with the real photographs of the Roman Pot and SSD package in Figs. 6.5 and 6.6.



Figure 6.5: Roman Pot vessel (photo).



Figure 6.6: Silicon Strip Detector packages stored in the protective atmosphere (photo).



Figure 6.7: Geant4 implementations of the Roman Pot vessel and SSD package with trigger counter.

6.3.3 Aperture tuning

It turned out during initial validations of the Geant4 simulation for run 15 with elastic proton-proton scattering events that the distribution of the (x, y) position of the proton in Roman Pots does not agree between the data and MC at a satisfactory level. It was understood that the perfect positioning of the elements of the collider

assumed in the geometry model may need some tuning, especially the positioning DX magnet which particularly limits acceptance of the RP detectors. The DX magnets are moved each time the switch between symmetric (e.g. $p + p$) and asymmetric (e.g. $p + Au$) collisions takes place to accommodate for the non-zero tilt of the beams in asymmetric collisions required to close beam orbits and provide collisions in the STAR IR. Therefore lateral offset of this element was expected to be the most significant ingredient needed to correctly describe forward proton acceptance in the RP location.

In order to quantitatively determine the agreement between the true DX position and position implemented in Geant4 a dedicated analysis of the DX “shadow” in the proton hit maps was performed. This algorithm looked for the sharp drop of event counts along the x coordinate in the histogram of y - vs. x -position of the track points in collinear elastic scattering events with single proton tracks on both sides of the IP. The result (for a single RP station) is presented in Fig. 6.8a and 6.8b for the data and MC, respectively. The envelope of the DX shadow found by the algorithm is marked with black circles. The points were transformed by swapping x and y axis and fitted with a circle, as shown in Fig. 6.8c. The parameters of the circle (radius R and center point (x_0, y_0)) do not reflect directly the position of the DX, but they were used to find the best shift of the

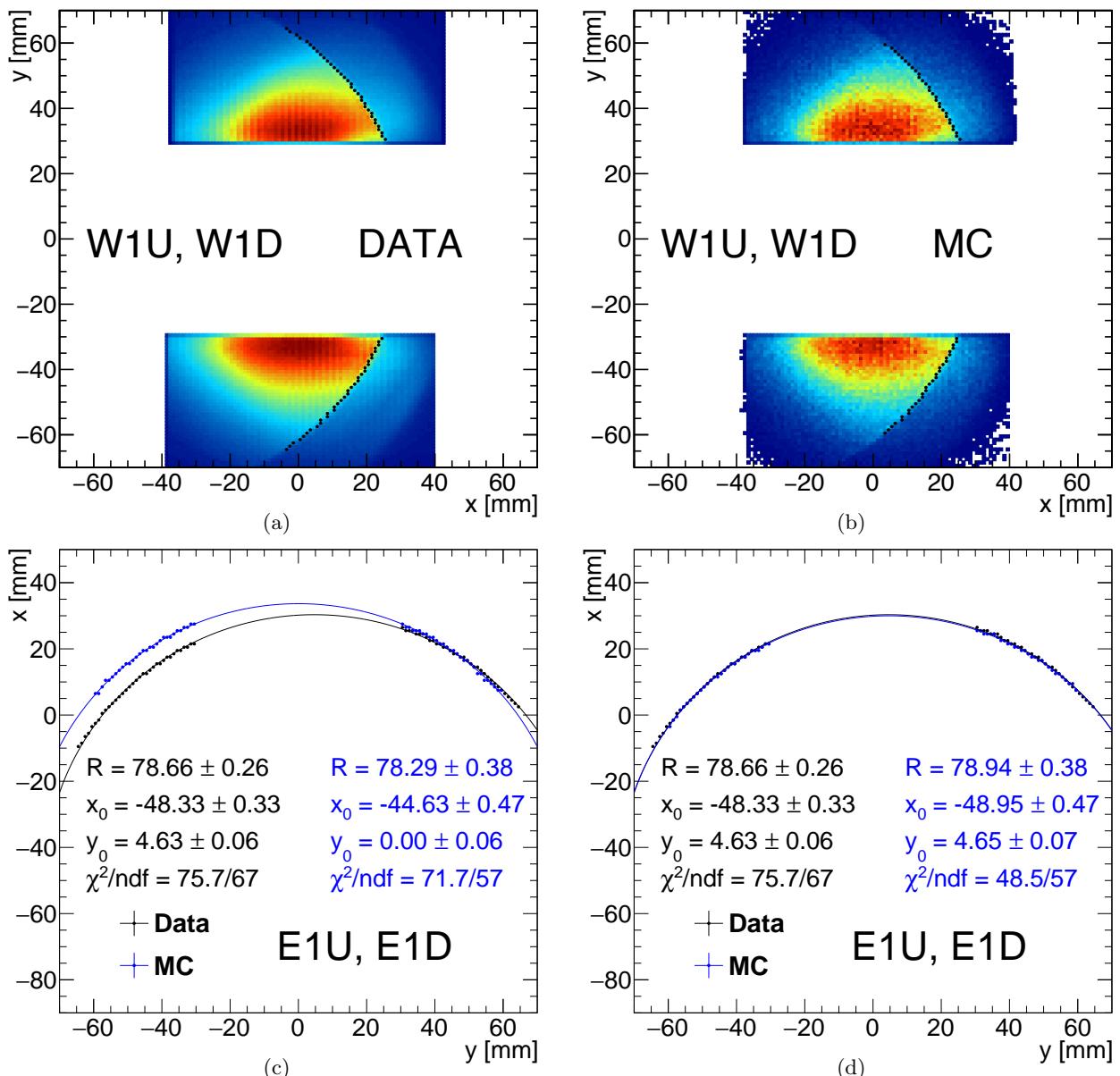


Figure 6.8: Sample hit map of track points of proton tracks in events selected as elastic scattering in the data (6.8a) and MC (6.8b) with the DX aperture "shadow" marked with black points. The DX aperture envelopes were fitted with circles (6.8c) which helped to establish the offsets of DX magnet positions in x and y with respect to the ideal (nominal) geometry, leading to nearly perfect agreement between the data and MC after introducing the offsets in Geant4 geometry (6.8d).

DX on the east and west side in the iterative method. The satisfactory agreement of the DX aperture envelopes between the data and MC was finally found when the shifts of the DX magnets with respect to their nominal positions were equal to $\Delta x_{\text{DX}}^{\text{East}} = -3.1$ mm, $\Delta y_{\text{DX}}^{\text{East}} = 4.0$ mm, $\Delta x_{\text{DX}}^{\text{West}} = -2.4$ mm, $\Delta y_{\text{DX}}^{\text{West}} = 0.4$ mm. The comparison of the DX shadow envelope after the tune is shown in Fig. 6.8d. All comparisons are presented in Appendix D.

6.3.4 Embedding technique

The pp2pp program has an option that allows to overlay the real data with the simulated detector response to MC events. The input data can be any trigger, but most common practice is to embed simulated signal into the zero bias triggers which by definition provide unbiased information about the real environment in which data were collected.

Merging of the simulated signal with the data signal is done for both the SSD data, as well as the PMT data. In case of the PMT data for each channel the ADC (energy) is set to the sum of values in the data and simulation, while the TAC (time) is set to larger value (earlier signal) of the two. In case of the SSD data the merging is done at the level of reconstructed clusters. At the end of simulation of an event a dedicated algorithm is run for every SSD plane which adds vectors of clusters from the data and from the simulation and merges overlapping clusters (recalculates their length, sums energy, updates cluster position). This new collection of clusters is saved in StMuRpsCollection together with new vectors of StMuRpsTrackPoints and StMuRpsTracks reconstructed from this new set of clusters.

7. dE/dx adjustment

Particle identification in our analyses is done using merged information from the TPC (specific energy loss of tracks dE/dx) and from the TOF (time of hit matched to TPC track). As can be seen in Fig. F.5, dE/dx information from the MC events simulated in STARsim (in red) poorly matches the data points (black). This results e.g. in large systematic error of estimate of particle identification efficiency.

This problem was discussed under ticket #3272 (Ref. [26]). There were trials to improve the TPC calibration in simulation, but the problem remained. It was finally concluded that the origin of the problem lies in the model of energy loss used in the STARsim, therefore any further action was postponed.

In order to tune simulated response of the TPC in terms of dE/dx , hence also reduce the systematic uncertainty related to particle identification, a correction method was developed based on proper transformation (recalculation) of simulated dE/dx to obtain new dE/dx whose distribution matches the data. We know that n_X^σ (where $X = \pi, K, p, \dots$) variable for particle X follows a gaussian distribution

$$n_X^\sigma = \left(\ln \frac{dE/dx}{\langle dE/dx \rangle_X} \right) / \sigma_{dE/dx}, \quad f(n_X^\sigma) = \mathcal{N}(n_X^\sigma; \mu = 0, \sigma = 1), \quad (7.1)$$

therefore dE/dx itself by definition follows log-normal distribution:

$$f(dE/dx) = \text{Log}\mathcal{N}(dE/dx; \mu = \langle dE/dx \rangle, \sigma = \sigma_{dE/dx}) = \frac{1}{\sqrt{2\pi} \cdot \sigma \cdot dE/dx} e^{-\frac{\ln^2 \frac{dE/dx}{\langle dE/dx \rangle}}{2\sigma^2}}. \quad (7.2)$$

The desired transformation should preserve the shape of dE/dx distribution so that it is still described by $\text{Log}\mathcal{N}$, however it should change μ and σ so that these values are equal to ones in the data. The transformation that satisfies above postulate is

$$dE/dx' = c \cdot (dE/dx)^a. \quad (7.3)$$

Parameters of the distribution $\text{Log}\mathcal{N}(dE/dx')$ are then

$$\mu' = c \cdot \mu^a, \quad \sigma' = a \cdot \sigma. \quad (7.4)$$

From above we get formulae for parameters of the transformation:

$$a = \sigma'/\sigma, \quad c = \frac{\mu'}{\mu^a}. \quad (7.5)$$

To sum up, one has to find the MPV and width parameter of the dE/dx spectrum of each particle in the data and MC, and use relations (7.5) in order to find parameters of the transformation introduced in Eq. (7.3).

The most challenging part of the task was extraction of the $\langle dE/dx \rangle$ and $\sigma_{dE/dx}$ from the data. In case of MC one can select tracks matched to true-level particles of given ID and thus separate dE/dx of different particles, which makes extraction of the distribution shape straightforward. Unfortunately, it is not possible to apply the same method to the data - here one has to deal with overlapping of the reconstructed dE/dx from different particles. Therefore fits of sum of $f(dE/dx)$ corresponding to different particles were performed to reconstructed track dE/dx in narrow momentum bins. The width of momentum bins was chosen to compromise statistics and validity of assumption of constant parameters of dE/dx distribution over bin range.

It was found during the fitting that log-normal distribution is not a perfect model of the reconstructed dE/dx . The problems with description of the data were mainly in the tail-part of the distribution from single particle. Precise model was necessary to obtain satisfactory quality of fits and trustworthy values of parameters. After some research the best model of dE/dx distribution from single particle was found to be

$$f(dE/dx) = \begin{cases} \frac{A}{\sqrt{2\pi} \cdot \sigma \cdot dE/dx} \exp \left(-\frac{1}{2} \left(\frac{\ln \frac{dE/dx}{\langle dE/dx \rangle}}{\sigma} \right)^2 \right) & \text{for } \frac{\ln \frac{dE/dx}{\langle dE/dx \rangle}}{\sigma} \leq k \\ \frac{A}{\sqrt{2\pi} \cdot \sigma \cdot dE/dx} \exp \left(-k \cdot \frac{\ln \frac{dE/dx}{\langle dE/dx \rangle}}{\sigma} + \frac{1}{2} k^2 \right) & \text{for } \frac{\ln \frac{dE/dx}{\langle dE/dx \rangle}}{\sigma} > k. \end{cases} \quad (7.6)$$

Such form was motivated by the function presented in Ref. [27], here adopted for the log-normal instead of normal distribution. Because the modification of the log-normal distribution is introduced only at high-end tail, the validity of the transformation discussed above still holds. To reduce fit complexity the k parameter was set the same for all particle species and fixed to value equal 2.2, which worked well for both data and embedded MC. Particles and their anti-particles were assumed to have the same dE/dx distributions for a given momentum and were analyzed together. The same track selection was used both for data and MC, with the quality criteria



Figure 7.1: Sample fit of sum of functions from Eq. (7.6) corresponding to different particle species to dE/dx spectra in the data in a single momentum bin.



Figure 7.2: Difference between MPV of dE/dx predicted by Bichsel parametrization and obtained from the fit of Eq. (7.6) to dE/dx distribution in the data (7.2c) and MC sample (7.2a) and dE/dx width parameter in data (7.2d) and MC (7.2b) as a function of reconstructed particle momentum for a few particle species. Solid lines represent fits to points of corresponding color. Only statistical errors are shown.

described in Sec. 2. The sample fit in a single momentum bin can be found in Fig. 7.1. Fits in all momentum bins can be found in Appendix F.

Results of the fits for all considered particle species (pions, kaons, protons, electrons and deuterons) are commonly presented in Fig. 7.2 with color markers. Figures 7.2c and 7.2a show the offset of the MPV of reconstructed dE/dx relative to the Bichsel parametrization in the data and embedded MC, respectively, and Fig. 7.2d and 7.2b show the width of reconstructed dE/dx (in the same order). Function able to qualitatively describe dependence of the parameters as a function of track momentum was empirically found to be given by Eq. (7.7):

$$g(p) = P_1 + P_2 \cdot \exp(-P_3 \cdot p) + P_4 \cdot \arctan(P_5 \cdot (p - P_6)) \quad (7.7)$$

This function was fitted to points corresponding to each particle type and fit result is shown in Fig. 7.2 with lines colored in accordance to markers. Values of parameters of above function are tabulated in Tab. 7.1.

PID	$\langle dE/dx \rangle_{\text{Bichsel}} - \langle dE/dx \rangle_{\text{MC}}$						$\sigma(dE/dx)_{\text{MC}}$					
	P_1	P_2	P_3	P_4	P_5	P_6	P_1	P_2	P_3	P_4	P_5	P_6
π^\pm	7.183e-8	-1.647e-4	41.68				0.0705			-1.42e-3	9.860	0.951
K^\pm	4.359e-8	-9.285e-6	7.697				0.0511	0.034	1.675	1.01e-2	4.934	0.528
p, \bar{p}	3.556e-8	-8.621e-6	3.980				0.0630	-7.725	27.17	3.37e-3	5.245	0.670
e^\pm	-6.219e-8	2.065e-7	3.241				0.0354	0.982	26.58	1.79e-2	41.515	0.095
d, \bar{d}	-1.305e-6	-5.268e-6	3.486				0.0967	-1526	18.75			

PID	$\langle dE/dx \rangle_{\text{Bichsel}} - \langle dE/dx \rangle_{\text{Data}}$						$\sigma(dE/dx)_{\text{Data}}$					
	P_1	P_2	P_3	P_4	P_5	P_6	P_1	P_2	P_3	P_4	P_5	P_6
π^\pm	-1.399e-8	2.012e-7	10.39				0.0734	1.907	31.86	-8.20e-4	22.788	0.653
K^\pm	2.325e-9	-3.690e-6	8.712				0.0808	-0.040	7.951	5.62e-3	-17.08	0.269
p, \bar{p}	-1.458e-7	0.6655	59.06	1.171e-7	4.660	0.644	0.0795	0.181	12.12			
e^\pm	9.005e-8	2.494e-7	8.834				0.0680	8.8e-4	1.549			
d, \bar{d}	-1.910e-7	5.637e-3	14.48				0.1161	-0.147	2.890			

(a)

(b)

Table 7.1: Parameters of functions from Fig. 7.2 describing reconstructed track dE/dx as a function of reconstructed momentum for a few particle species in the data (7.1b) and embedded MC (7.1a). Blank cells denote parameters equal 0. Units of parameters P_i are such that if one provides momentum in Eq. (7.7) in GeV/c the resultant offset of dE/dx MPV with respect to Bichsel parametrization is in GeV/cm , and the resultant σ parameter is unitless.

The correctness of the entire procedure described in this section was verified by comparing the reconstructed track dE/dx between the data and embedded MC without and with the dE/dx transformed using Eq. (7.3) and parameters from Tab. 7.1. Some difficulty arised in this comparison due to inconsistent relative content of different particle species in the data and embedded MC sample. Problem was ressolved by separating dE/dx histograms of different particle species in MC (in the same way as it was done for extraction of dE/dx MPV and σ for each particle ID) and fitting the sum of histograms from different particle types to the data histogram (in momentum bins). The only free parameters in the fit were relative contents of histogram from singe particle type in the data histogram. A sample comparison between the dE/dx in data and embedded MC is presented in Fig. 7.3. Comparison in all other momentum bins is contained in Appendix F (Fig. F.5). Fits were done for adjusted dE/dx (filled green). Histograms for unadjusted dE/dx (hashed red) were composed using the same relative content of particles as obtained from the fit of adjusted dE/dx . The ratio of the MC to the data shown in the lower pad of Fig. 7.3 and Fig. F.5 clearly demonstrates better agreement of the MC and the data in terms of position and width of peaks in dE/dx spectrum after the adjustment.

Exactly the same cross-check was done for n_X^σ variables, which are directly used to identify particles in our analyses. They carry information about the distance (in width units) of reconstructed dE/dx from the MPV of dE/dx according to Bichsel parametrization - they are kind of a pull variables. For every track n_X^σ was recalculated using its definition as shown in Eq. (7.8):

$$n_X^\sigma = \ln \left(\frac{dE/dx'}{\langle dE/dx \rangle_{\text{Bichsel}}^X} \right) / \delta(dE/dx), \quad (7.8)$$

where instead of original dE/dx the adjusted one (dE/dx') was used. Uncertainty of reconstructed dE/dx , $\delta(dE/dx)$, was left unchanged. The sample comparison of n_{pion}^σ , n_{kaon}^σ and n_{proton}^σ variables can be found in Fig. 7.4. Comparison in full set of momentum bins is placed in Appendix F. As in case of dE/dx comparison,

in the lower pads of Fig. 7.4 the ratio of the MC to the data also demonstrates better agreement of the MC and the data, which was the goal of the adjustment.



Figure 7.3: Sample comparison of dE/dx spectrum between data and embedded MC (before and after dE/dx adjustment) in a single momentum bin. Lower pad shows the ratio between embedded MC and data before and after dE/dx adjustment. In both upper and lower pads the same color code is used. Only statistical errors are shown. Due to limited statistics of embedded MC some data points do not have corresponding entries in MC.



Figure 7.4: Sample comparison of n_{pion}^σ (7.4a), n_{kaon}^σ (7.4b) and n_{proton}^σ (7.4c) distribution between data and embedded MC in single momentum bin. Lower pads show the ratio between embedded MC and data before and after dE/dx adjustment. In both upper and lower pads the same color code is used. Only statistical errors are shown.

8. TPC track pointing resolution adjustment

It was found during the analysis that distributions of quantities which describe the pointing resolution of the TPC tracks do not agree well between the data and embedded MC. Namely, the resolutions of the global helices associated with the tracks were found to be significantly better in the STAR simulation than in the data, what manifests as narrower DCA and d_0 distribution in the embedded MC, comparing to corresponding distribution in the data (Fig. 8.5). This issue was discussed under ticket #3332 (Ref. [28]).

This problem could affect the momentum resolution and thus all other resolutions and response matrices used in data unfolding. Therefore the resolution adjustment procedure was performed to find appropriate parameters of the “artificial” helix deterioration and finally obtain agreement between DCA and d_0 distributions (and all related resolutions) in the data and embedded MC.

In order to worsen pointing resolution in the MC a smearing of the helix radius $\sigma(R)$ was introduced. Based on d_0 comparison in Fig. 8.5a it was decided to account also for the small systematic bias of the helix radius $\Delta\mu(R)$ ¹, which may be present e.g. due to differences in the material budget used the simulation and reconstruction. Both smearing and bias of the helix radius were introduced only for MC tracks which were matched with the true-level particles since only simulated tracks require adjustment (tracks from zero-bias event used in embedding already contain all detector effects).

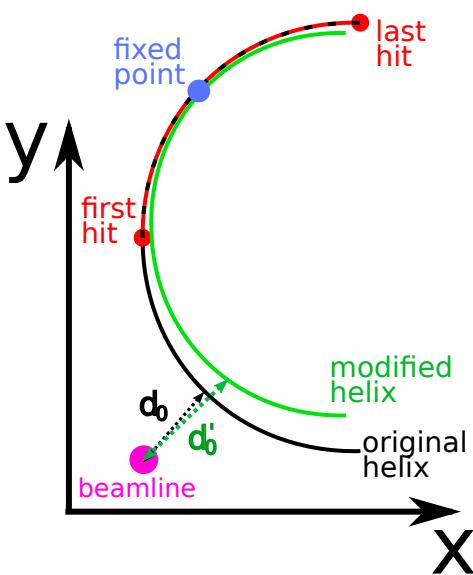


Figure 8.1: Sketch of helix modification procedure and d_0 calculation.

Extraction of $\Delta\mu(R)$ and $\sigma(R)$ parameter required to achieve agreement of pointing resolution between embedded MC and the data involved a few steps, as listed below:

1. Series of d_0 histograms in bins of p_T (100 MeV/c wide) was prepared, each for different size of distortion (different $\Delta\mu(R)$ and $\sigma(R)$) of global helix of the TPC tracks matched with true-level particles (example plot in single p_T bin is shown in Fig. 8.2):

- (a) for each set of parameters $\Delta\mu(R)$ and $\sigma(R)$ the helix radius R was recalculated independently for each track following the Eq. (8.1):

$$R' = R \times \mathcal{N}(1 + \Delta\mu(R), \sigma(R)), \quad (8.1)$$

- (b) new helix of a radius R' was assigned to a track and used to calculate d_0 . The modified helix was obtained by changing the radius of original helix from R to R' with a fixed middle point between the first and last TPC hit of a global track represented by the helix (Fig. 8.1). The momentum of the track was also recalculated:

$$p'_T = p_T \times \frac{R'}{R}, \quad \eta' = \eta \times \frac{R'}{R}. \quad (8.2)$$

2. In each p_T bin the χ^2/NDF was calculated between the data and MC d_0 histogram in a range $-1.5 \text{ cm} < d_0 < 1.5 \text{ cm}$ (corresponding to d_0 cut used in analyses), for every point in parameter space of radius distortion (for every set of $\Delta\mu(R)$ and $\sigma(R)$). An example (single p_T bin) of map of $-\chi^2/\text{NDF}$ in a parameter space is presented in Fig 8.3.
3. In each bin of recalculated p_T the 2-dim parabola $z(x, y; a, b, x_0, y_0, z_0)$ given in Eq. (8.3) ($z = \chi^2/\text{NDF}$, $x = \Delta\mu(R)$, $y = \sigma(R)$) was fitted to $-\chi^2/\text{NDF}$ in the global minimum region to obtain the best-fit distortion parameters.

$$z = z_0 - a(x - x_0)^2 - b(y - y_0)^2. \quad (8.3)$$
4. The best-fit smearing $\sigma(R)$ (equal to parabola parameter y_0) and best-fit bias $\Delta\mu(R)$ (x_0) from individual p_T bins was plotted as a function of global track p_T (Fig. 8.4). Each point was assigned with an error being a quadratic sum of two components: the error on x_0 (y_0) resulting from the parabola fit to $-\chi^2/\text{NDF}$, and length of corresponding semi-axis of ellipsis formed by the intersection of fitted parabola with the xy -plane at $z = z_0 - 1/\text{NDF}$ (from definition of the parameter uncertainty given by the change of overall χ^2 by 1 unit). Resultant formulae for the error of each individual point in Fig. 8.4 are

¹Transverse impact parameter d_0 takes positive value if the beamline is contained inside the helix (in the yz -plane projection), otherwise it is negative. Any asymmetry in the d_0 distribution in the MC with respect to the data indicates presence of systematic difference in reconstructed d_0 , hence also in reconstructed R .



Figure 8.2: Example of comparison of d_0 histograms in single p_T bin in the data (black points) and embedded MC (colored lines) in the procedure of TPC pointing resolution adjustment. Only MC histograms for $\Delta\mu(R) = 0$ and $\sigma(R) = 0, 5 \times 10^{-3}$ and 10^{-2} were shown for explanatory purposes.

$$\delta(\Delta\mu(R)) = \sqrt{\delta_{\text{fit}}^2(x_0) + \frac{1}{2a\text{NDF}}}, \quad (8.4)$$

From Fig. 8.2 one can read that $\text{NDF} = 14$. In calculation of uncertainties correlation of $\Delta\mu(R)$ and $\sigma(R)$ have not been accounted.

5. The empirically determined functions were fitted to points representing $\Delta\mu(R)$ and $\sigma(R)$ dependence on the global track p_T . Their form and values of parameters are given in Fig. 8.4.

Helices of global TPC tracks were deteriorated according to Eq. (8.1) and the parametrizations of global track p_T -dependence of $\Delta\mu(R)$ and $\sigma(R)$ from Fig. 8.4, to verify if better agreement between the data and embedded MC is found after the adjustment. Filled histograms in Fig. 8.5 show d_0 and DCA distributions after the described adjustment, and filled circles in the bottom pad show their ratio to the data points. Clearly, there is much better agreement between embedded MC and the data after the pointing resolution adjustment. Remaining differences may arise from incomplete theoretical model of the CEP process implemented in GenEx leading to different p_T spectra of the data and the model (e.g. model does not contain resonant $\pi^+\pi^-$ production).

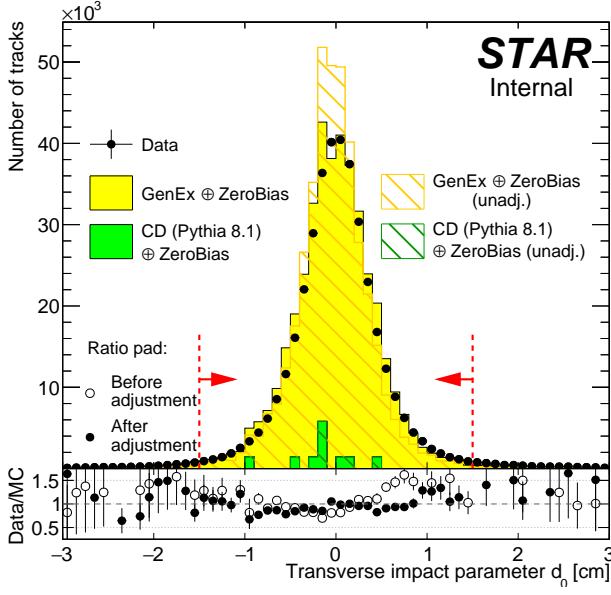


Figure 8.3: Example of $-\chi^2/\text{NDF}$ map in a parameter space in the procedure of TPC pointing resolution adjustment. The red surface represents parabola fitted in the vicinity of the global minimum.

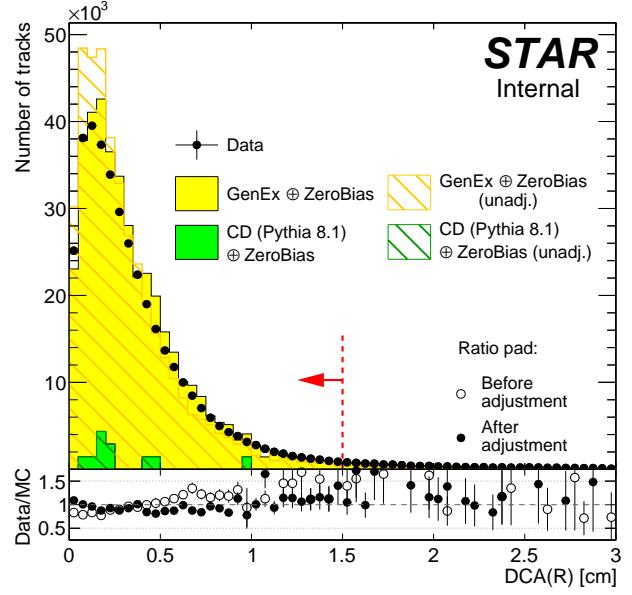
$$\delta(\sigma(R)) = \sqrt{\delta_{\text{fit}}^2(y_0) + \frac{1}{2b\text{NDF}}}. \quad (8.5)$$



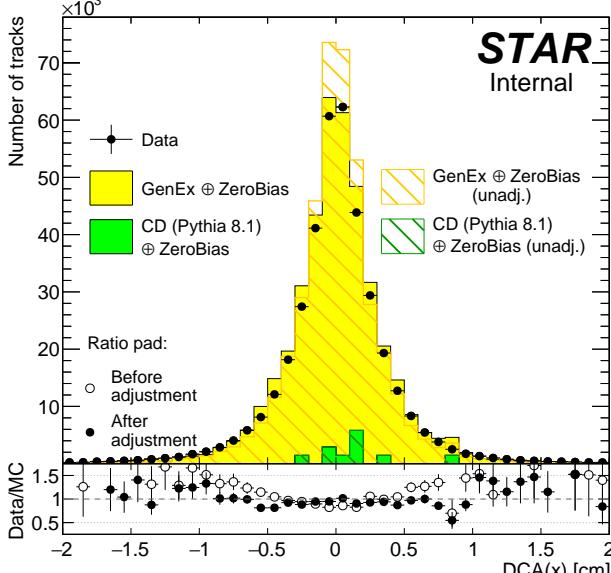
Figure 8.4: Best-fit parameters obtained in the procedure of the TPC track pointing resolution adjustment. Uncertainties on parameters resulting solely from the fit of Eq. (8.3) to $-\chi^2/\text{NDF}$ are represented by the lines with perpendicular endings. Total uncertainties (Eqs. (8.4), (8.5)) extend beyond. The empirical functions fitted to points are drawn with corresponding colors, and formula of each is written aside.



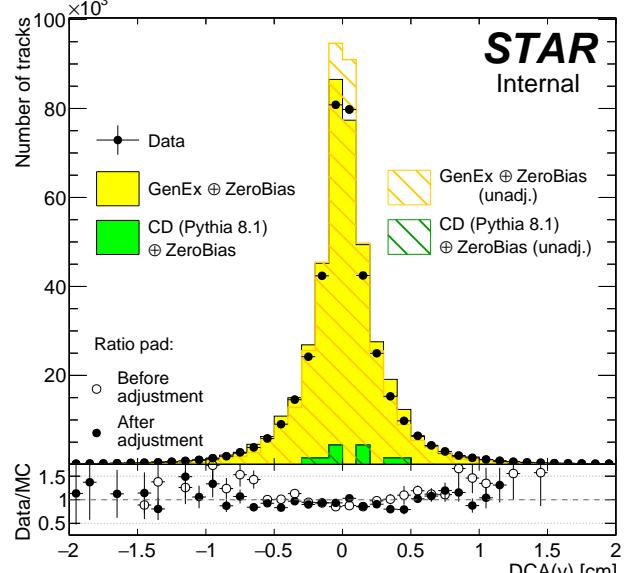
(a)



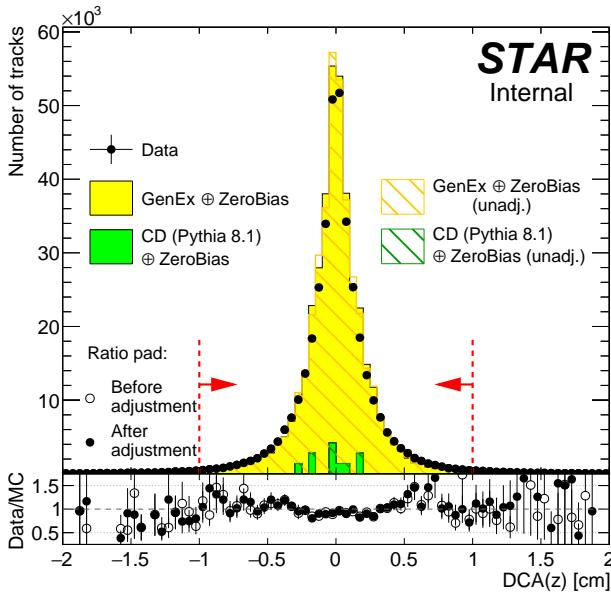
(b)



(c)



(d)



(e)

Figure 8.5: Comparison of distribution of pion transverse impact parameter d_0 (8.5a) and transverse (8.5b), x - (8.5c), y - (8.5d) and z -component (8.5e) of the DCA vector between the global helix and primary vertex in the data (CEP) and embedded MC (GenEx). Distributions for unadjusted helices are drawn as hashed histograms, while filled histograms are for adjusted helices. Normalizations of the signal and backgrounds were established from the comparison of p_T^{miss} and $\Delta\theta$ distributions after full selection (without cut on the presented quantity and without exclusivity cut), as described in Sec. XXX of Ref. [1]. Red dashed lines and red arrows indicate the range of each quantity which is accepted in analyses.

9. Dead material in front of TPC

Particle detected and reconstructed in the TPC must first pass through the detector material standing in between the accelerator vacuum and TPC gas. This affects track reconstruction efficiency, as the particle may interact with that material - in worst case inelastically, and induce secondary particles thus lower reconstruction efficiency. Accuracy of modeling of the detector material in the STAR simulation, especially in run 15 with the HFT installed, influences systematic error e.g. on the TPC track reconstruction efficiency. In this section the density of secondary vertices is compared between the data and embedded MC. The density of secondary vertices is directly proportional to the amount of the material in given volume, hence any discrepancy between secondary vertex distribution in the data and MC can be a hint for inaccuracies of the STAR simulation which should be accordingly covered by the systematic uncertainties. It should be stressed that this analysis is not aimed to tune the material budget in the STAR simulation, as there are much better data for this than high-luminosity proton-proton collisions from run 15. The aim of presented study is to obtain reasonable estimate of the component of systematic uncertainty of the TPC track reconstruction efficiency related to the error on the amount and distribution of inactive material.

Analysis of the distribution of secondary vertices was performed using both zero-bias (ZB) data and minimum-bias MC (Pythia) embedded into zero-bias triggers. Because of insufficient statistics of the ZB data, for the purpose of analysis presented in this section both standard ZB data sample (from ZB triggers in st_rp stream) and the subsample of RP_CP triggers (see Ref. [7] for trigger details) with identified elastic proton-proton scat-

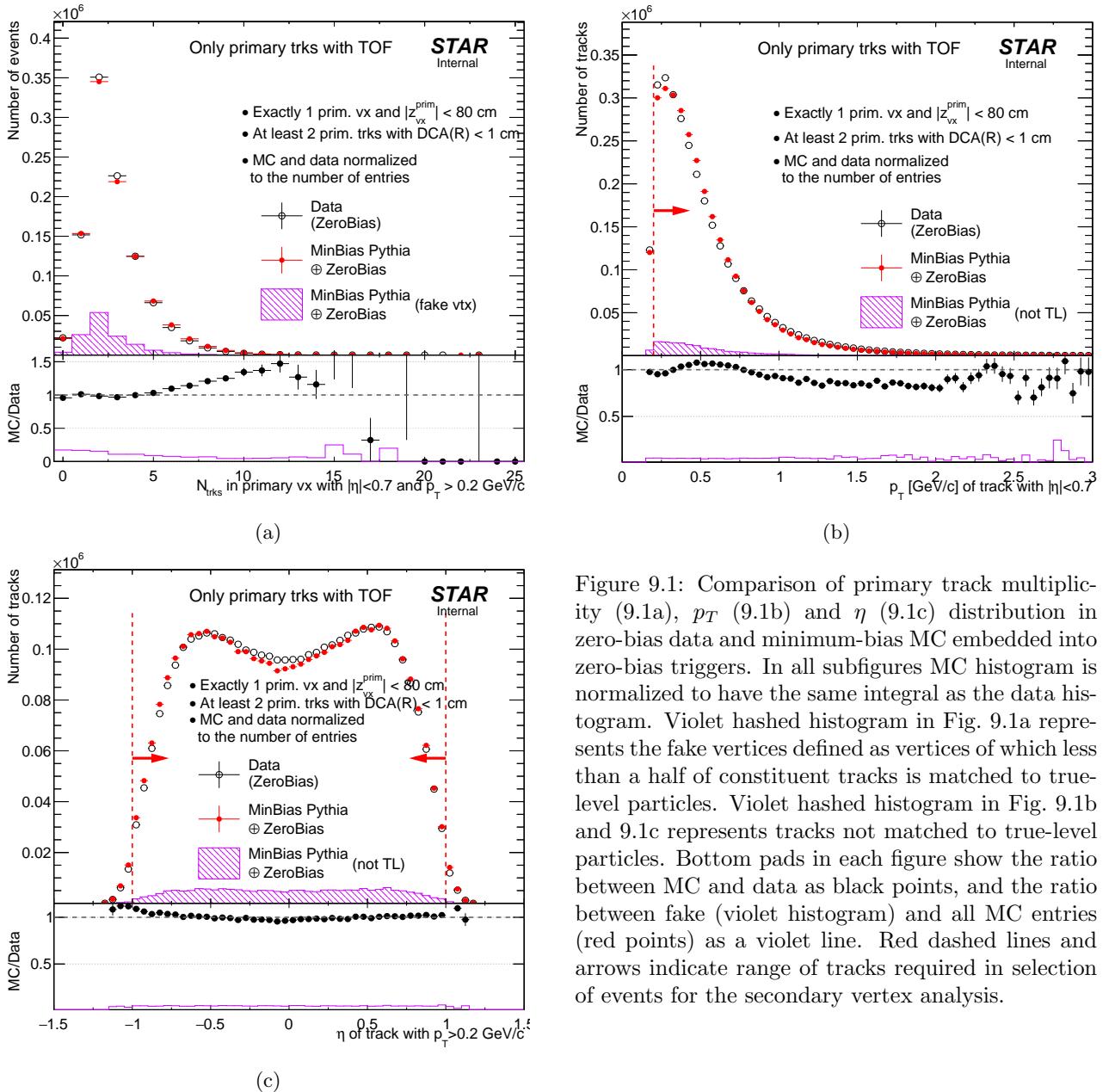


Figure 9.1: Comparison of primary track multiplicity (9.1a), p_T (9.1b) and η (9.1c) distribution in zero-bias data and minimum-bias MC embedded into zero-bias triggers. In all subfigures MC histogram is normalized to have the same integral as the data histogram. Violet hashed histogram in Fig. 9.1a represents the fake vertices defined as vertices of which less than a half of constituent tracks is matched to true-level particles. Violet hashed histogram in Fig. 9.1b and 9.1c represents tracks not matched to true-level particles. Bottom pads in each figure show the ratio between MC and data as black points, and the ratio between fake (violet histogram) and all MC entries (red points) as a violet line. Red dashed lines and arrows indicate range of tracks required in selection of events for the secondary vertex analysis.

tering events using loose RP track selection were used. The latter subsample is in good approximation a ZB sample in terms of central detector, as it was triggered only by the east and west coincidence of Roman Pots - any particle present in the TPC and TOF is predominantly a product of pile-up interaction. In all plots and later in the text we refer to this merged sample as ZB data sample.

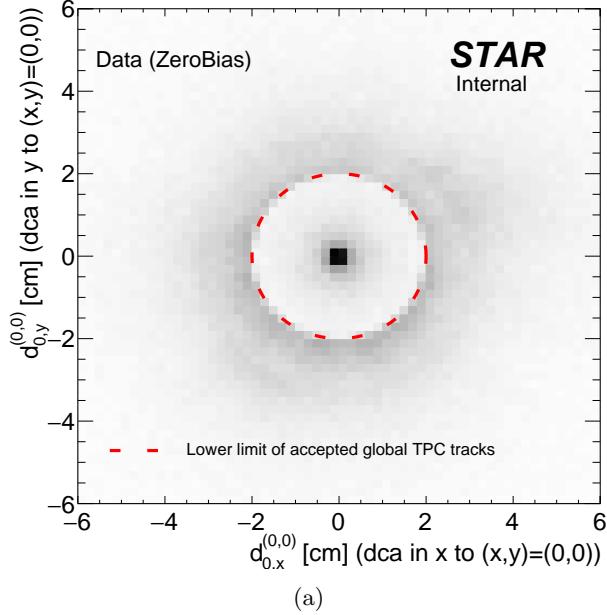
Analysis started with the following selection of events:

1. Exactly 1 reconstructed primary vertex (with tracks matched to hits in TOF; beamline constraint was used in reconstruction therefore primary vertices lie on the beamline and tracks associated with them have global DCA (d_0) not larger than ≈ 1.5 cm),
2. $|z_{vx}| < 80$ cm,
3. ≥ 2 prim. TOF tracks with: $DCA(R) < 1$ cm, $|\eta| < 1$, $p_T > 0.2$ GeV/c, $N_{\text{hits}}^{\text{fit}} \geq 25$, $N_{\text{hits}}^{\text{dE/dx}} \geq 15$.

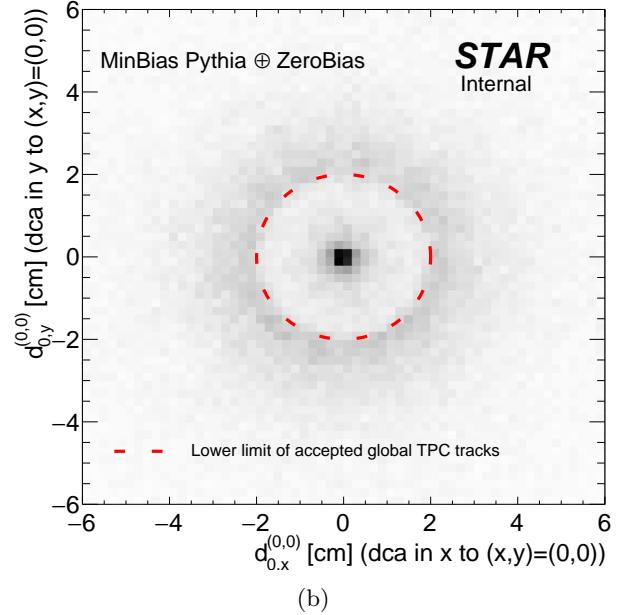
The aim of above criteria was to select pile-up-free events with well defined vertex. Cut on z -vertex is identical to one used in physics analyses. Figure 9.1 shows comparison of quantities characterizing an event. In general a moderate agreement between MC and data can be observed, considered sufficient for trustworthy result of described analysis.

As a next step the TPC tracks were selected for the search and reconstruction of secondary vertices. The requirements were as follows:

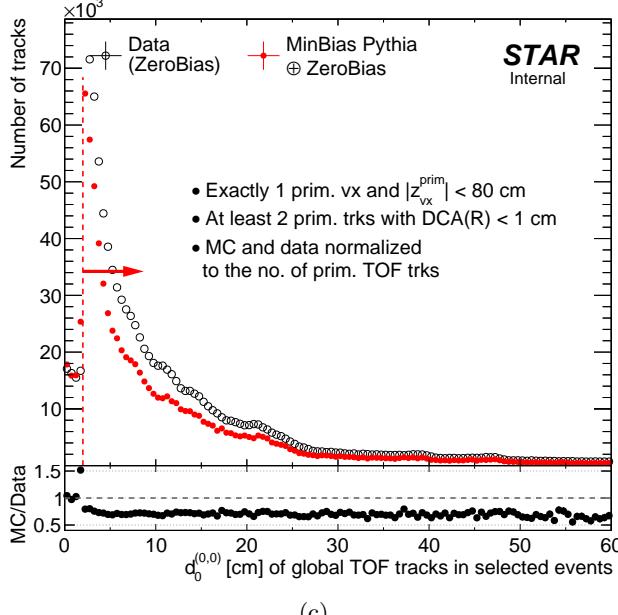
1. Global TPC tracks matched with TOF not associated with any primary TPC track,
2. $|\eta| < 0.7$, $p_T > 0.2$ GeV/c, $N_{\text{hits}}^{\text{fit}} \geq 25$, $N_{\text{hits}}^{\text{dE/dx}} \geq 15$, $N_{\text{hits}}^{\text{fit}}/N_{\text{hits}}^{\text{poss}} \geq 0.52$,
3. Distance of closest approach to the STAR z -axis ($x, y = (0, 0)$, $d_0^{(0,0)}$), larger than inner radius of the beampipe: $d_0^{(0,0)} > 2$ cm.



(a)



(b)



(c)

Figure 9.2: Two-dimensional $d_0^{(0,0)}$ distribution of global TPC tracks matched with TOF in zero-bias data (9.2a) and embedded minimum-bias MC (9.2b), and comparison of their radial projection in wider range of $d_0^{(0,0)}$ (9.2c). Red dashed lines and arrow indicate limit of $d_0^{(0,0)}$ of tracks accepted for the secondary vertex analyses (limit equals 2 cm). In all subfigures there are only entries from global tracks not associated with the primary tracks. Even with relatively low pointing resolution of the TPC tracks (~ 1 cm) one can recognize structures which can be attributed the beampipe starting at $d_0^{(0,0)} = 2$ cm, and HFT elements at about 8 cm, 11 cm, 14 cm and 22 cm.

These cuts were intended to select in-time TPC tracks with high chance of being a product of secondary interaction of primary particle with the detector material. The higher limit of accepted $d_0^{(0,0)}$ was set in analysis, the less background (fake secondary vertices) was found in the secondary vertex distribution for a price of limited access to secondary vertices of low radial distance from STAR z -axis. Cut of 2 cm was found a good compromise. In Fig. 9.2 we present comparison of $d_0^{(0,0)}$ distribution of selected global TOF-matched TPC tracks in the data and embedded MC (without cut on $d_0^{(0,0)}$). Number of secondary vertices is proportional to both material density and flux of primary particles. To remove bias due to the different fluxes of primary particles in data and simulation the latter was scaled by the following factor:

$$\text{MC normalization factor} = \frac{\langle N_{\text{trks/evt}}^{\text{DATA}} \rangle \times N_{\text{evts}}^{\text{DATA}}}{\langle N_{\text{trks/evt}}^{\text{MC}} \rangle \times N_{\text{evts}}^{\text{MC}}} = \frac{N_{\text{trks}}^{\text{DATA}}}{N_{\text{trks}}^{\text{MC}}} \quad (9.1)$$

Especially in Fig. 9.2c one can find structures/peaks that might be attributed to subdetectors (PXL, IST, SST) of the HFT. Notable is different yield of histograms which on first glance indicate different amount of simulated dead material with respect to real conditions. However, the main reason for this difference in yields was found in imperfect simulation of the pointing resolution of the TPC tracks. As shown in Chap. 8 the track pointing resolution is better in the STAR simulation comparing to the data, therefore in MC more true primary tracks are reconstructed as primary tracks (are forming/attached to the primary vertices), hence less such tracks are accepted in the selection of global track for secondary vertex reconstruction (comparing to data). The pointing resolution worsening which is introduced to MC tracks (see Chap. 8) does not help in this case because the track smearing is introduced after primary vertices are reconstructed (after MuDst are produced). This effect is accounted later in the background subtraction procedure to reveal the studied difference in the amount of existing and simulated dead material.

After secondary track candidates were selected, the following algorithm for secondary vertex reconstruction was used:

1. Loop over all pairs of secondary track candidates, store pairs whose DCA is less than 0.5 cm (nearby tracks passing a proximity cut),
2. Link pairs of nearby tracks into sets of tracks connected by the common nearby tracks,
3. Loop over all sets defined in 2., in each set loop over all pairs from given set, reject worst-matching tracks (these with largest DCA to others) until all pairs of tracks have DCA less than 0.5 cm,
4. Based on number of tracks in secondary vertex, total charge, specific energy loss, dE/dx , cosine of the opening angle of two tracks $\cos(\Delta\theta)$ and invariant mass of two tracks m_{inv} determine if the vertex is from resonance decay, photoconversion or is of nuclear/hadronic nature:
 - (a) if ≥ 2 tracks in the vertex, or 2 tracks of the same sign, or 2 tracks of the opposite sign and cosine of the opening angle between two tracks $\cos(\Delta\theta) < -0.99$ (tracks back-to-back) vertex is recognized as hadronic; otherwise (sub-sample of 2 opposite-sign tracks case remains)
 - (b) if $|n_{\text{pion}}^\sigma| < 3$ for both tracks and invariant mass (assuming pion mass) is within $[0.470, 0.515]$ GeV (Fig. 9.4a) the vertex is recognized as originating from the resonance decay (K_S^0); otherwise
 - (c) if $|n_{\text{pion}}^\sigma| < 3$ for one track and $|n_{\text{proton}}^\sigma| < 3$ for the other, and invariant mass (assuming pion and proton mass, respectively) is within $[1.107, 1.123]$ GeV (Fig. 9.4b) the vertex is recognized as originating from the resonance decay ($\Lambda, \bar{\Lambda}$); otherwise
 - (d) if $|n_{\text{electron}}^\sigma| < 3$ for both tracks and invariant mass (assuming electron mass) is less than 0.09 GeV and cosine of the opening angle between two tracks $\cos(\Delta\theta) > 0.95$ (Fig. 9.4c) the vertex is recognized as originating from photoconversion; otherwise the vertex is recognized as hadronic.
5. Calculate the vertex position as the average DCA point of all track pairs in the vertex.

As a result secondary vertices were reconstructed, whose multiplicity distribution is depicted in Fig. 9.3. Analysis was continued only with vertices of multiplicity equal 2. The first reason was that most of vertices consist of just a pair of tracks. Another reason was the background subtraction method developed only for vertices made of two tracks. In addition to this, only vertices representing primary particles in the pseudorapidity range $-0.7 < \eta < 0.7$ were analyzed. To enable such selection a variable η_{vtx} was defined, as shown in Fig. 9.5.



Figure 9.4: Cosine of the opening angle between two opposite-sign tracks recognized as e^+e^- pair (9.4c) and invariant mass of two opposite-sign tracks recognized as $\pi^+\pi^-$ (9.4a) and $\pi^+\bar{p}/\pi^-\bar{p}$ pair (9.4b) forming a secondary vertex. Dashed red lines and red arrows indicate ranges of quantities which define particular nature of the secondary vertex (photoconversion/resonance decay/hadronic). All three plots represent zero-bias data.

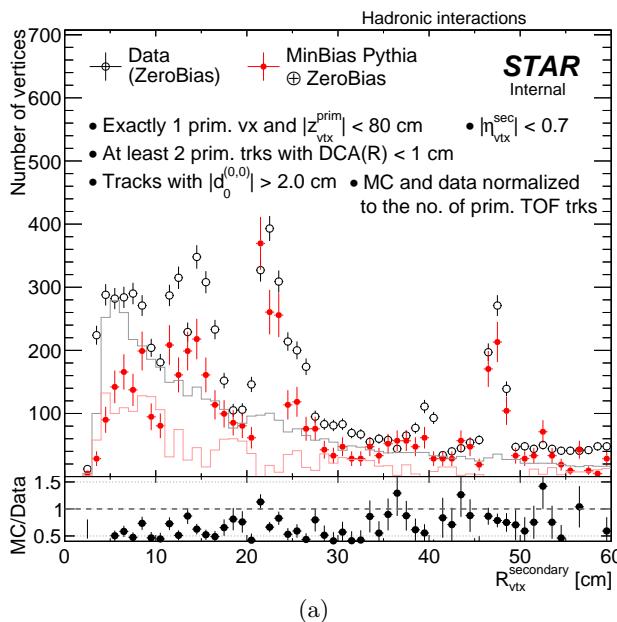


Figure 9.5: η_{vtx} definition (sketch).

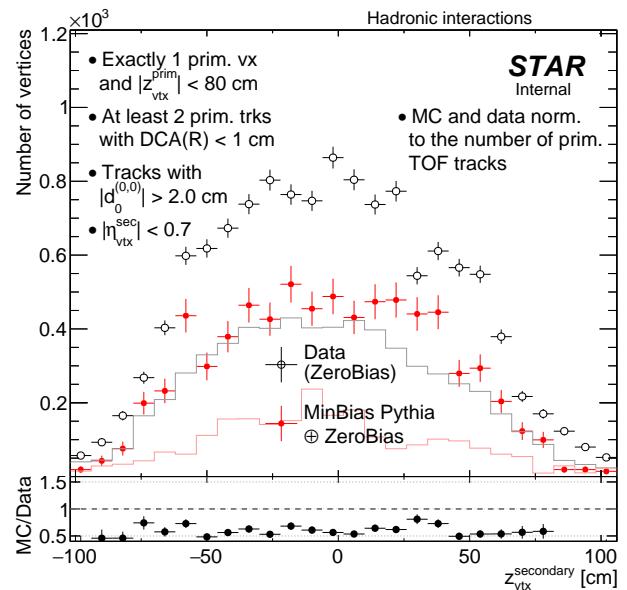
(fake pairs) distributed over the distance of closest approach between two tracks. A comment needs to be made that the agreement of the shape of the tails in data and MC distributions was achieved only after the adjustment of the TPC resolution in MC, as described in Sec. 8. This agreement allows to believe in proper description of the

Raw distributions of $R_{\text{vtx}}^{\text{secondary}}$ and $z_{\text{vtx}}^{\text{secondary}}$ are shown in Fig. 9.6a and Fig. 9.6b, respectively. In $R_{\text{vtx}}^{\text{secondary}}$ spectrum one can find peaks in the regions where the HFT subdetectors are placed. Peaks seem to lie on top of a tail whose origin has been identified with the secondary vertices made of pairs containing true primary tracks which were not associated with any primary vertex and unfortunately passed selection of global tracks for the secondary vertex reconstruction. Without this background subtracted, the ratio of MC to data varies mostly between 0.5 and 0.7. For this reason a method of estimation of the background was invented, as described in the next paragraph.

Background estimation makes use of different content of fake secondary vertices depending on the proximity cut used in the secondary vertex reconstruction. Figure 9.7 shows the percentage of background



(a)



(b)

Figure 9.6: Comparison of raw $R_{\text{vtx}}^{\text{secondary}}$ (9.6a) and $z_{\text{vtx}}^{\text{secondary}}$ (9.6b) distribution in the data (opened black circles) and embedded MC (filled red circles). Only vertices recognized as products of hadronic interactions are shown in the figure. Solid lines denote estimated background content in the distribution of corresponding color.



Figure 9.7: Comparison of DCA between all pairs of secondary track candidates selected for the secondary vertex reconstruction in the data and embedded MC. MC histogram is normalized to the data at $DCA > 5$ cm. Violet hashed histogram depicts pairs contained in MC histogram and not originating from the same vertex. Solid violet line in the lower pad denotes ratio of violet and red histogram.

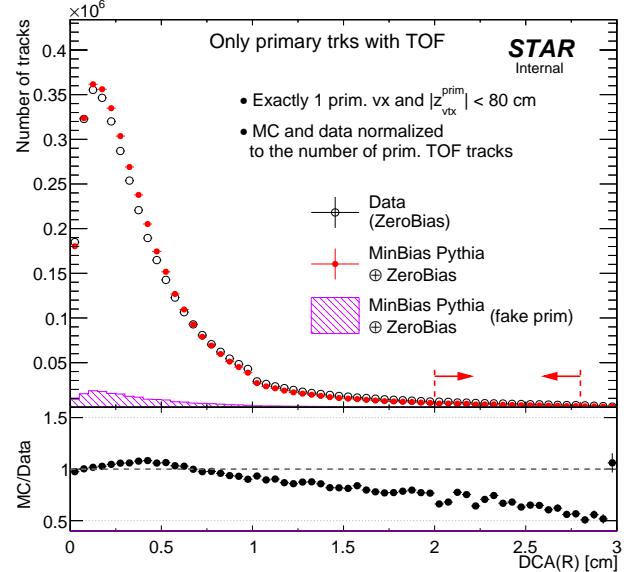


Figure 9.8: Comparison of radial DCA of all primary tracks matched with TOF and passing quality criteria in events selected for secondary vertex analysis, between the data and embedded MC. Violet hashed histogram represents tracks not matched to true-level particles. Red dashed lines and arrows limit region used to find normalization that compensates different background yield in reconstructed secondary vertex distributions in data and embedded MC. Discontinuity at 1 cm is due to selection of events with at least 2 tracks of $DCA(R) < 1$ cm

data by MC in terms of background distribution over DCA of two global tracks, which is used in the background estimation.

It agrees with intuition that the most optimal cut to select pairs from the secondary vertices is as low as about 0.5 cm, however one can select sample with slightly different ratio of signal to background if the proximity cut is changed to accept tracks of DCA within some higher limits. In Fig. 9.7 the nominal proximity cut is marked with the green line (signal region), while the modified proximity cut is marked with red lines (background region). With such two versions of cuts used in vertexing the two independent distributions of secondary vertices can be obtained: one with the standard proximity cut - \mathcal{H}_1 , the other with modified proximity cut, in our case $1.0 \text{ cm} < DCA < 2.5 \text{ cm}$ - \mathcal{H}_2 . Limits in modified proximity cut were set to such values in order to ensure enough statistics as well as provide satisfactory resolution of secondary vertex position calculated as a middle point between DCA points on helices associated with the tracks. One can note that the content of histograms can be described by the set of equations given below:

$$\left\{ \begin{array}{l} \mathcal{H}_1 = (1 - B) \times \text{signal} + B \times \text{background}, \\ \mathcal{H}_2 = (1 - B') \times \text{signal} + B' \times \text{background}, \end{array} \right. \quad (9.2)$$

$$\left\{ \begin{array}{l} \text{signal} = \frac{B' \times \mathcal{H}_1 - B \times \mathcal{H}_2}{B' - B} \\ \text{background} = \frac{(1 - B) \times \mathcal{H}_2 - (1 - B') \times \mathcal{H}_1}{B' - B} \end{array} \right. \quad (9.3)$$

in which parameters B and B' denote the background fraction in the distribution resultant from analysis utilizing nominal and modified proximity cut, respectively. The solution to set of Eqs. (9.2), (9.3) is the following:

$$\left\{ \begin{array}{l} \text{signal} = \frac{B' \times \mathcal{H}_1 - B \times \mathcal{H}_2}{B' - B} \\ \text{background} = \frac{(1 - B) \times \mathcal{H}_2 - (1 - B') \times \mathcal{H}_1}{B' - B} \end{array} \right. \quad (9.4)$$

$$\left\{ \begin{array}{l} \text{signal} = \frac{B' \times \mathcal{H}_1 - B \times \mathcal{H}_2}{B' - B} \\ \text{background} = \frac{(1 - B) \times \mathcal{H}_2 - (1 - B') \times \mathcal{H}_1}{B' - B} \end{array} \right. \quad (9.5)$$

An important remark here is that the background fraction extracted from the ratio of violet and red histograms in Fig. 9.7 can be used directly in Eqs. (9.2)-(9.5) only for background estimation in MC. In case of background estimation in data parameters B and B' have to be corrected for the “leakage” of true primary tracks to set of selected secondary track candidates due to worse TPC track pointing resolution in the data and thus lower probability/efficiency of attaching the true primary tracks to primary vertices at large $DCA(R)$, as it

was described in one of preceding paragraphs. The correction factor κ is extracted from the ratio of the radial DCA of the primary TPC tracks in events selected for the secondary vertex study (Fig. 9.8). This ratio at large DCA(R) provides estimate of how many more true primary TPC tracks in the data is not recognized as primary tracks and hence overpopulate/enrich sample of global tracks which are selected for the reconstruction of primary vertices, by definition increasing number of fake, background secondary vertices. Histogram range selected for calculation of the ratio was set to $2.0 \text{ cm} < \text{DCA}(R) < 2.6 \text{ cm}$, as this range coincides with the $d_0^{(0,0)}$ of global tracks accepted for the analysis. κ calculated in this range equals 1.48. Variation of value of κ with changed limits of DCA(R) selected for the ratio calculation do not influence significantly the final result. The correction is done by multiplying fraction B and B' by κ only when estimating the background in the data.

Background determined with the described method is shown in Fig. 9.1 with the solid lines colored according to corresponding markers. This background was subtracted and resulting distributions of the secondary vertex positions in the transverse and longitudinal direction are presented in Fig. 9.9. Most relevant region - the HFT detector extending between $\sim 2 \text{ cm}$ and $\sim 30 \text{ cm}$ is satisfactorily well described by MC. Also, the inner wall of the TPC at $\sim 48 \text{ cm}$ well matches between data and MC. Looking at the ratios in the bottom panels of Figs. 9.1 one can conclude that the differences between the dead material density in the data and STAR simulation generally vary between $\pm 25\%$, which we consider a systematic uncertainty on the amount of the simulated dead material in front of TPC. Related systematic uncertainty on the TPC track reconstruction efficiency is described in Sec. 10.1.2.



Figure 9.9: Comparison of background-subtracted $R_{\text{vtx}}^{\text{secondary}}$ (9.9a) and $z_{\text{vtx}}^{\text{secondary}}$ (9.9b) distribution in the data (open black circles) and embedded MC (filled red circles). Only vertices recognized as products of hadronic interactions are shown in the figure.

10. Systematic errors

10.1 TPC track reconstruction efficiency

10.1.1 Embedding (pile-up) effect

One major difference between simulation and real data is the presence of pile-up events. The average number of pile-up tracks in a triggering event is proportional to the BBC coincidence rate. It is expected that the difference between simulation and real data drops at lower BBC rates, and the effects of pile-up tracks could be much reduced by fitting the tracking efficiency as a function of BBC rate and using the extrapolated value at zero luminosity to compare with simulation.

The embedded MC was divided into two samples due to mean BBC_AND rate: $\langle \text{BBC_AND} \rangle = 700 \text{ kHz}$ and $\langle \text{BBC_AND} \rangle = 1400 \text{ kHz}$. Next, the track reconstruction efficiency was calculated for those two samples and no-pile-up MC corresponding to them. The difference between TPC track reconstruction efficiencies for pile-up and no-pile-up MCs was calculated as:

$$\Delta\epsilon_{TPC}^{1400/700 \text{ kHz}} = \frac{N_{reco}^{no-pile-up} - N_{reco}^{pile-up}}{N_{gen}} \quad (10.1)$$

where:

N_{gen} -number of MC tracks,

$N_{reco}^{no-pile-up}$ - number of reconstructed tracks, matched with MC tracks in no-pile-up MC,

$N_{reco}^{pile-up}$ - number of reconstructed tracks, matched with MC tracks in pile-up MC.

The difference between high and low pile-up runs is given by:

$$\Delta\epsilon_{TPC} = \Delta\epsilon_{TPC}^{1400 \text{ kHz}} - 2 \cdot \Delta\epsilon_{TPC}^{700 \text{ kHz}} \quad (10.2)$$

Finally, above difference, shown in Figs. 10.2 and 10.3 for π^\pm , varies between 2–3% and was taken as systematic uncertainty related to TPC track reconstruction efficiency.

Figure 10.2: π^\pm TPC track reconstruction efficiency as a function of p_T ($|\eta| < 0.7$, $|V_z| < 80 \text{ cm}$) for embedded MC samples with $\langle \text{BBC_AND} \rangle \geq 700 \text{ kHz}$ and $\langle \text{BBC_AND} \rangle \geq 1400 \text{ kHz}$. The efficiencies from corresponding no-pile-up MC samples were also shown. Additionally, the differences from Eq. 10.2 were drawn in the bottom of each plot.

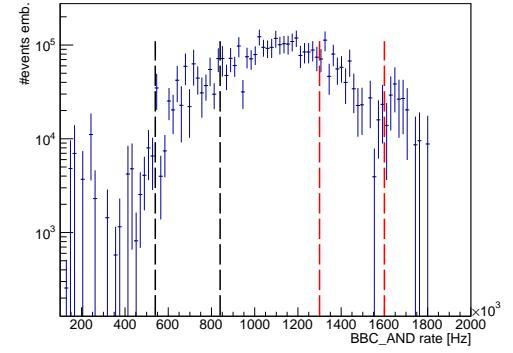
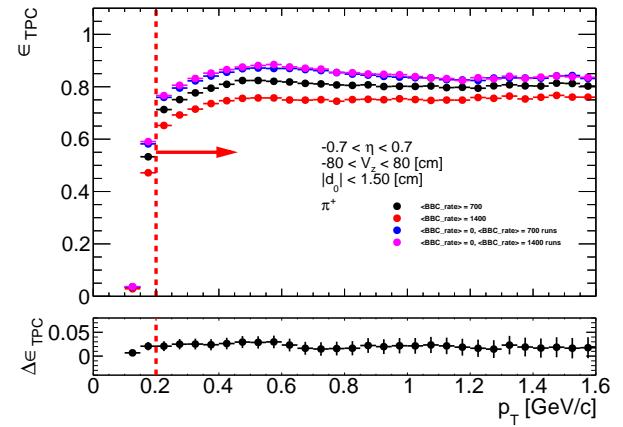
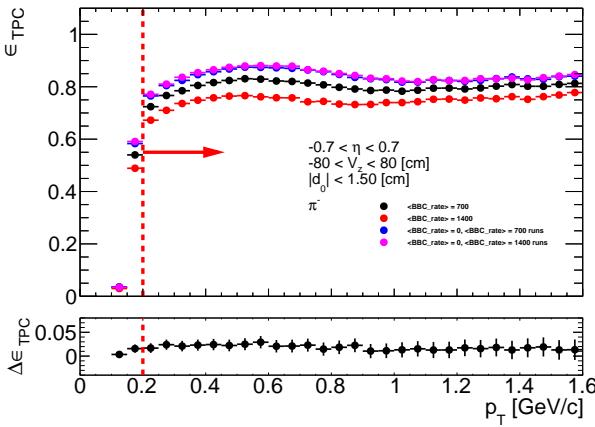
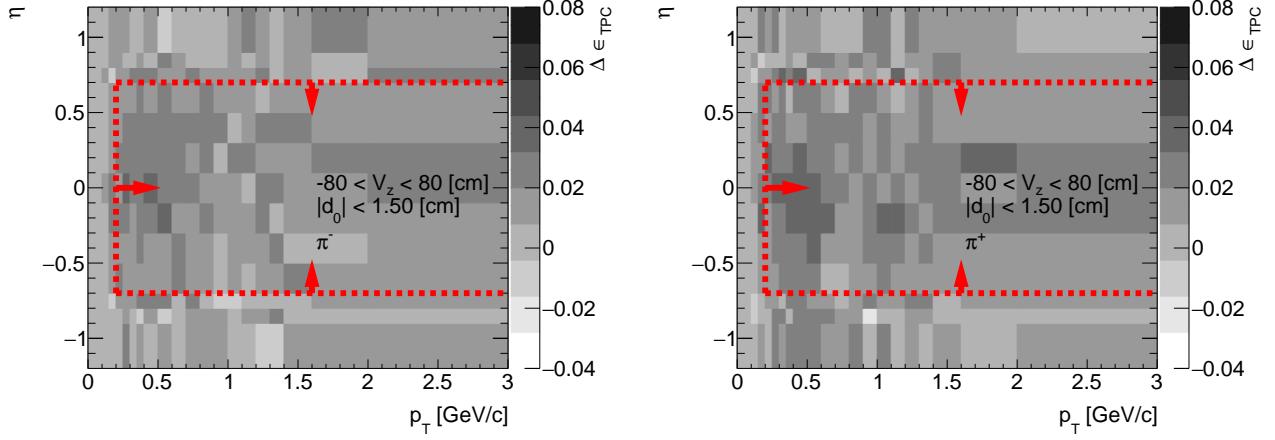


Figure 10.1: Number of events in embedded MC as a function of BBC_AND rate. The black and red lines represent the events with $\langle \text{BBC_AND} \rangle \geq 700 \text{ kHz}$ and $\langle \text{BBC_AND} \rangle \geq 1400 \text{ kHz}$, respectively.

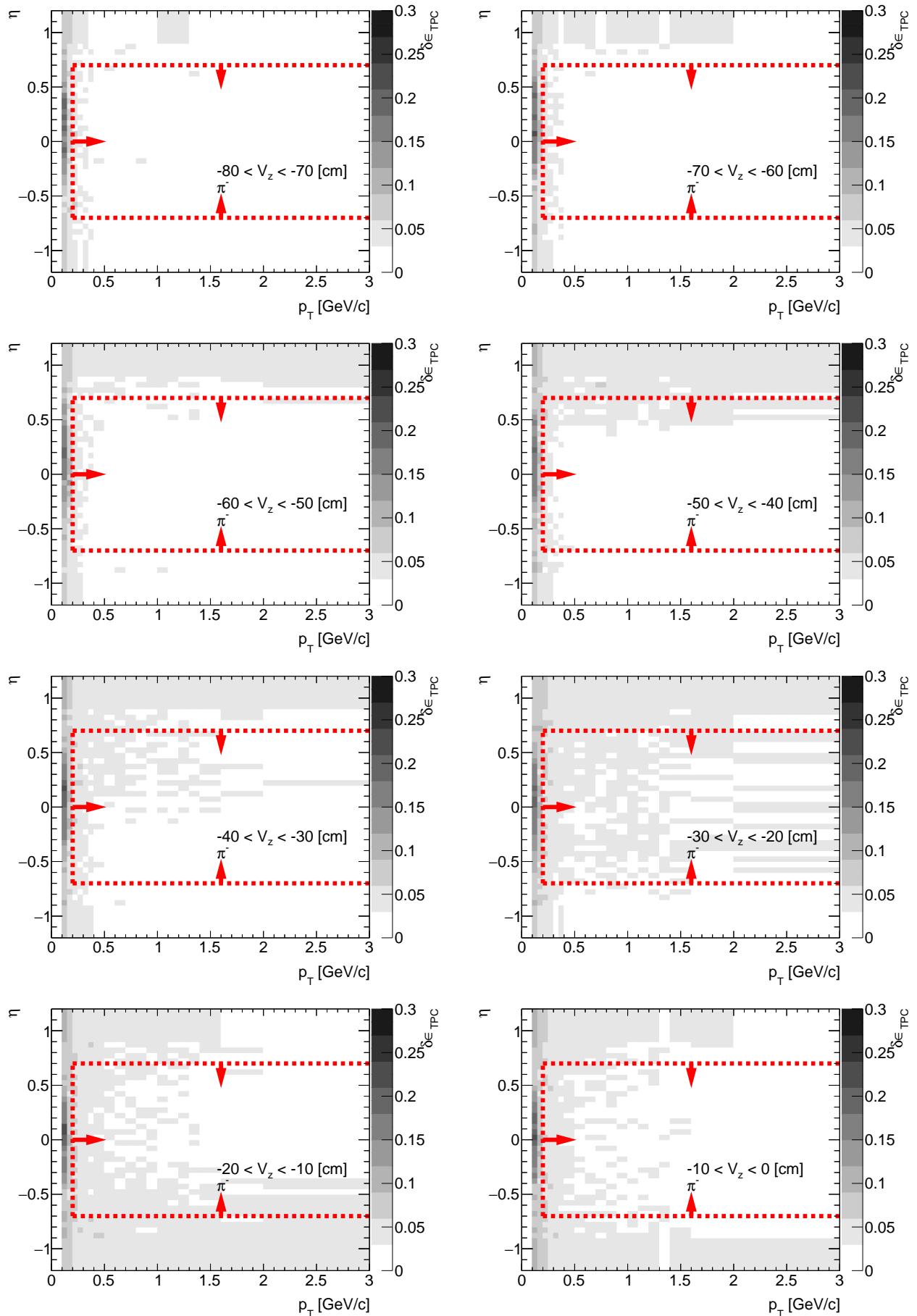
Figure 10.3: The difference $\Delta\epsilon_{TPC} = \Delta\epsilon_{TPC}^{1400 \text{ kHz}} - 2 \cdot \Delta\epsilon_{TPC}^{700 \text{ kHz}}$ for π^\pm as a function of p_T and η ($|V_z| < 80 \text{ cm}$).



10.1.2 Dead material effect on TPC track reconstruction efficiency

The amount of dead material in front of TPC differs up to 20% between data and simulation (see Sec. 9). First, the amount of lost particles, $\delta\epsilon_{TPC}$, due to the interaction with dead material in front of TPC was estimated using no-pile-up MC samples. The results for π^- in CD are shown in Fig. 10.4. Then the symmetric systematic uncertainty to the TPC track reconstruction efficiency due to dead material was introduced as $\pm 0.2 \cdot \delta\epsilon_{TPC}$. In Fig. 10.5 the systematic uncertainty is shown for each particle species in CD as a function of p_T ($|\eta| < 0.7, |V_z| < 80 \text{ cm}$). The results for other particles and SD are shown in Figs. in Appendix E.

Figure 10.4: The amount of lost π^- due to the interaction with dead material in front of TPC. Each plot represents the fraction of lost π^- , $\delta\epsilon_{TPC}$ (z-axis), as a function of true particle pseudorapidity η (y-axis) and transverse momentum p_T (x-axis) in single z -vertex bin.



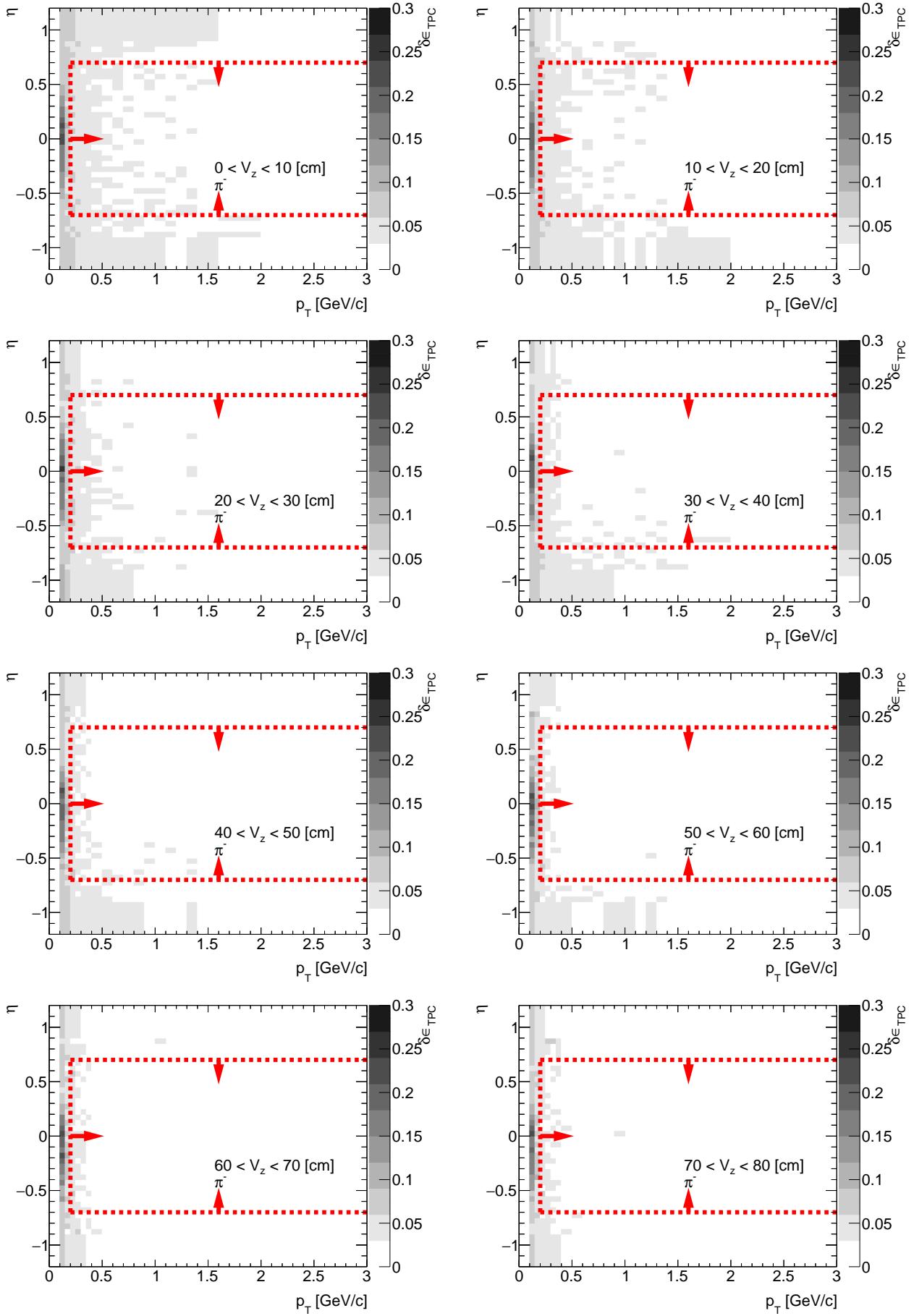
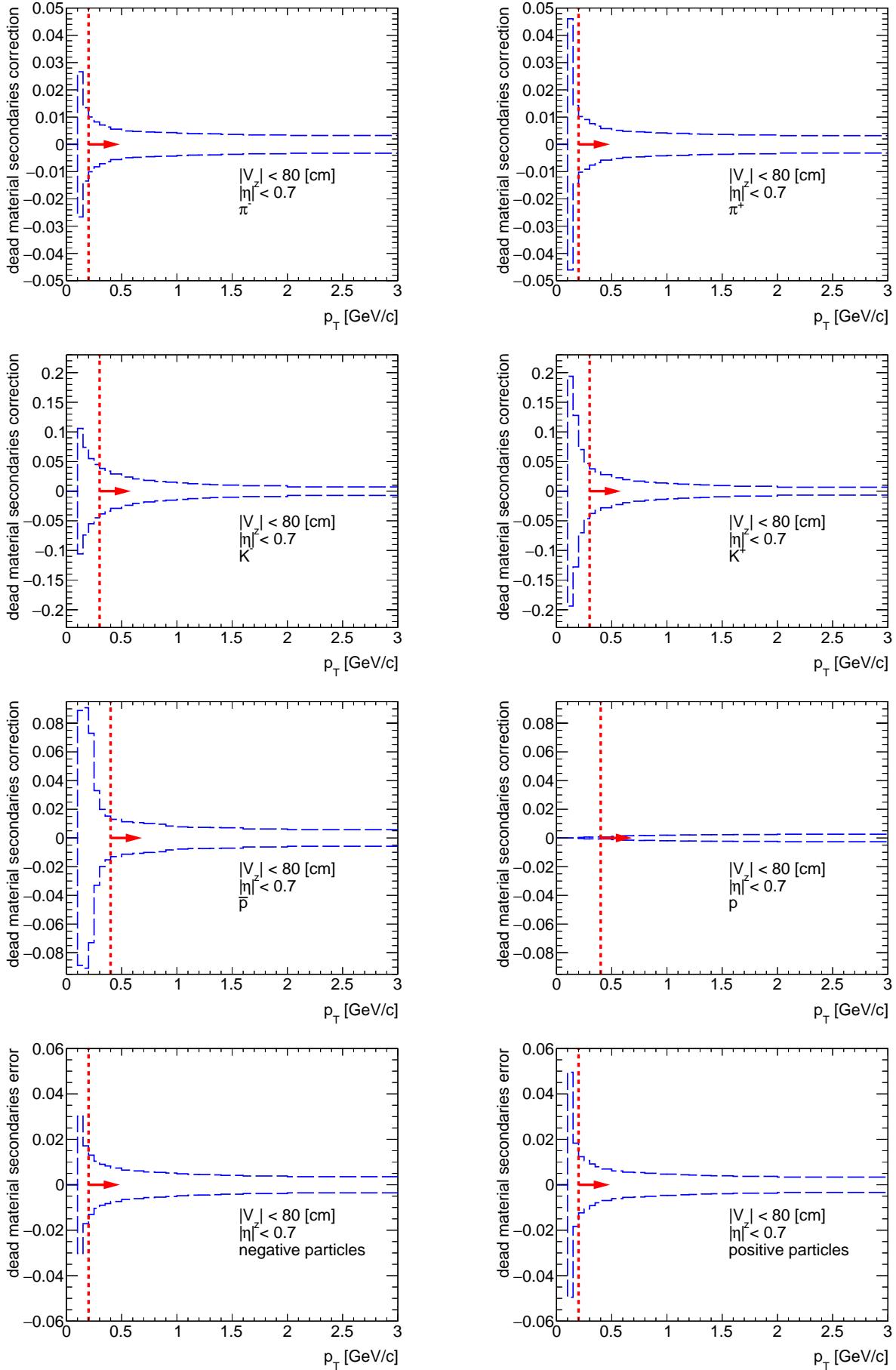


Figure 10.5: The systematic uncertainty to the TPC track reconstruction efficiency due to amount of dead material in front of TPC using MC samples for CD. Each plot represents the systematic uncertainty as a function of true particle p_T ($|\eta| < 0.7, |V_z| < 80 \text{ cm}$) for given particle species: $\pi^-, \pi^+, K^-, K^+, \bar{p}$ and p . It was also calculated for negative and positive particles without identification.



10.2 TOF matching efficiency

10.2.1 Embedding (pile-up) effect

The approach to calculate the systematic uncertainty on TOF matching efficiency related to pile-up was quite similar to the one used for TPC track reconstruction efficiency (Sec. 10.1.1). However, the TOF matching efficiency is conditional and depends on TPC track reconstruction efficiency. Since that, the difference between high and low pile-up runs is given by:

$$\Delta\epsilon_{TOF}^{1400/700 \text{ kHz}} = \frac{N_{TPC-TOF}^{no-pile-up}}{N_{TPC}^{no-pile-up}} - \frac{N_{TPC-TOF}^{pile-up}}{N_{TPC}^{pile-up}} \quad (10.3)$$

where:

$N_{TPC-TOF}^{pile-up}$ - number of reconstructed tracks, matched with MC tracks and TOF hit in pile-up MC,

$N_{TPC-TOF}^{no-pile-up}$ - number of reconstructed tracks, matched with MC tracks and TOF hit in no-pile-up MC,

$N_{TPC}^{pile-up}$ - number of reconstructed tracks, matched with MC tracks in pile-up MC,

$N_{TPC}^{no-pile-up}$ - number of reconstructed tracks, matched with MC tracks in no-pile-up MC.

Next the difference between high and low pile-up events was calculated with the formula similar to the one given by Eq. 10.2 and is shown in Figs. 10.6 and 10.7. The origin of $N_{TPC-TOF}$ increase is not known (it may be due to lack of pile-up in TPC or TOF). Since that, it is impossible to correctly calculate the statistical error for $\Delta\epsilon_{TOF}$. Nevertheless, $\Delta\epsilon_{TOF}$ is smaller than 0.5% and can be neglected in comparison with other systematic uncertainties.

Figure 10.6: π^\pm TOF matching efficiency as a function of p_T ($|\eta| < 0.7$, $|V_z| < 80$ cm) for embedded MC samples with $\langle BBC_AND \rangle \geq 700$ kHz and $\langle BBC_AND \rangle \geq 1400$ kHz. The efficiencies from corresponding no-pile-up MC samples were also shown. Additionally, the differences from Eq. 10.2 were drawn in the bottom of each plot.

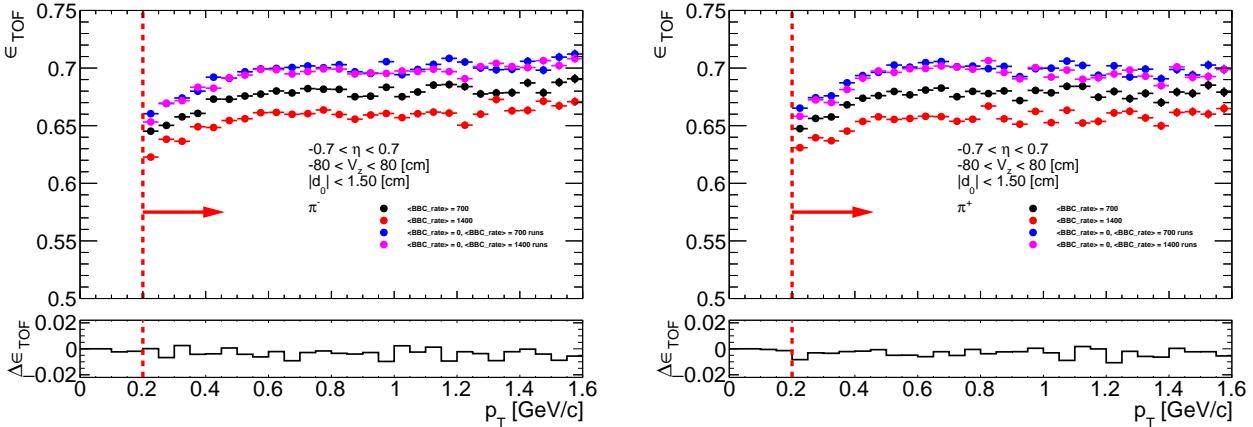
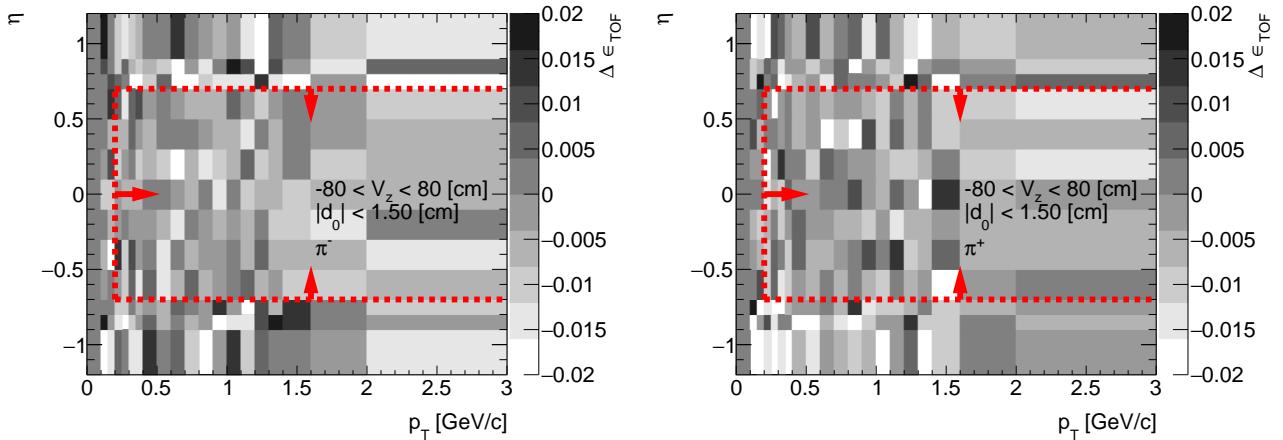


Figure 10.7: The difference $\Delta\epsilon_{TOF} = \Delta\epsilon_{TOF}^{1400 \text{ kHz}} - 2 \cdot \Delta\epsilon_{TOF}^{700 \text{ kHz}}$ for π^\pm as a function of p_T and η ($|V_z| < 80$ cm).



10.2.2 Absolute error on TOF efficiency

Systematic uncertainty of the TOF efficiency related to the accuracy of the TOF system simulation in STARsim and the TOF efficiency correction derived in Sec. 4.1 was estimated by comparing that nominal TOF efficiency with the one obtained with an independent method described below.

In some STAR analyses the TOF hit reconstruction and matching efficiency is determined from the data with the use of BEMC: real (in-time) tracks are selected based on the fact that they match to BEMC cluster. If they do, the TOF efficiency is calculated as a ratio of number of TOF-matched tracks to number of all tracks. This solution may provide slightly biased efficiency, because the signal in the detector placed behind TOF, such as BEMC, ensures that particle still followed the original helical path past the last hit of the track in TPC (Fig. 10.8).

We decided to calculate the TOF efficiency using the TPC tracks containing hits in HFT....

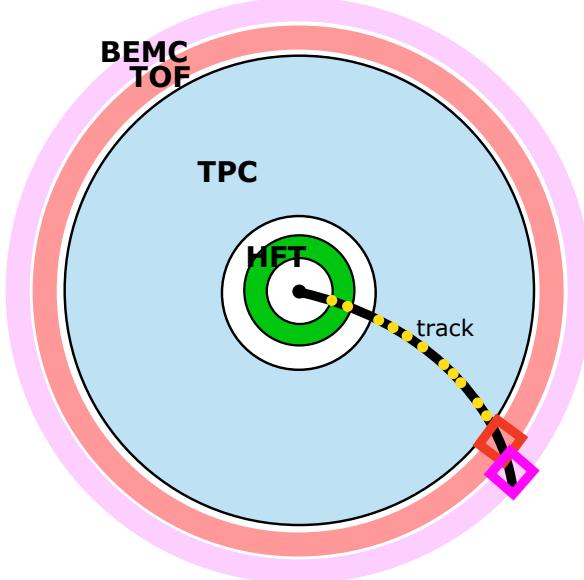


Figure 10.8: Sketch of the cross section of the central detector and the track reconstructed with points in HFT. Presence of HFT points in a reconstructed track can be used as a tagger of the in-time tracks.

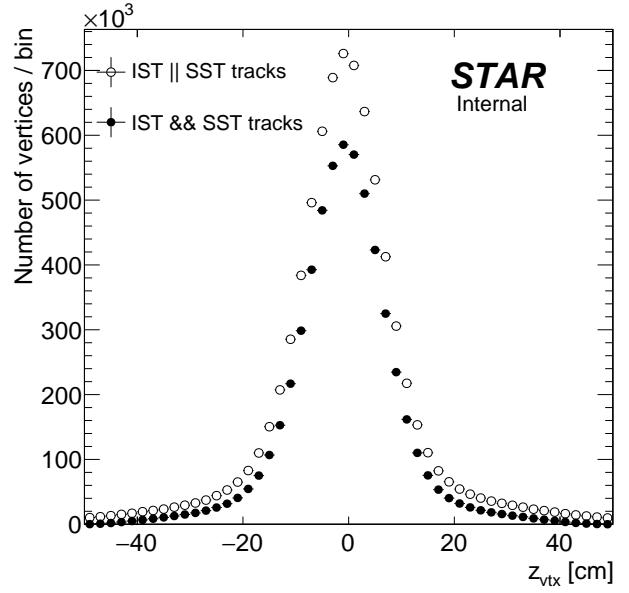
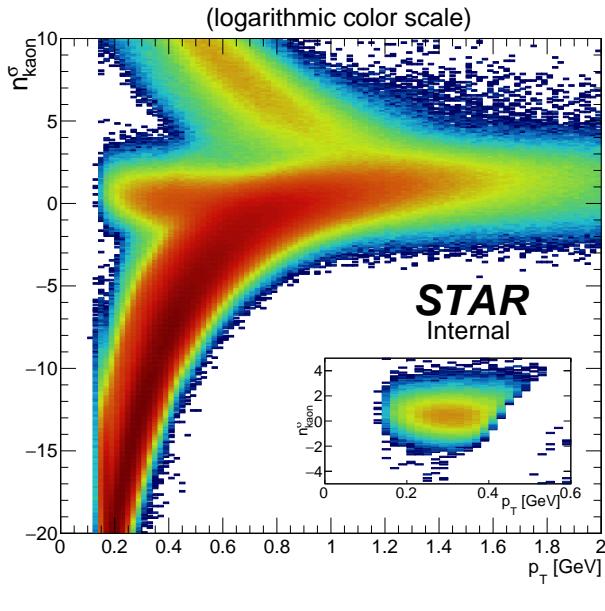
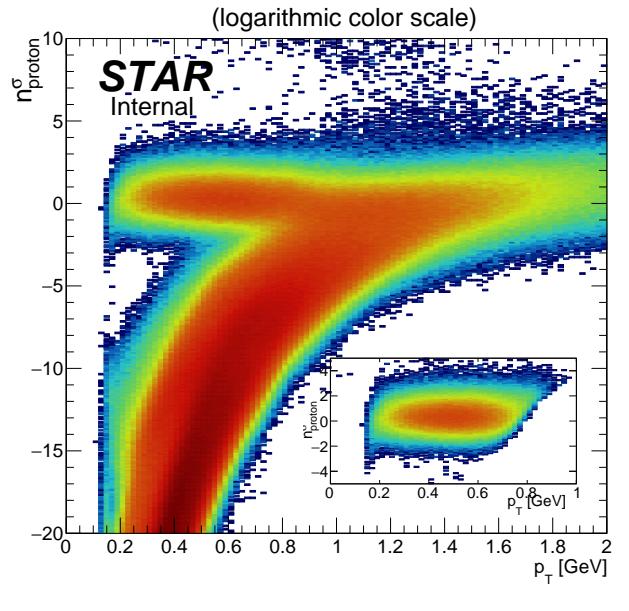


Figure 10.9: Distribution of z -position of vertices containing TPC tracks with HFT hits (st_ssd stream). Open circles represent vertices with tracks with hits in IST or SST, full circles - IST and SST.



(a)



(b)

Figure 10.10: Casase.

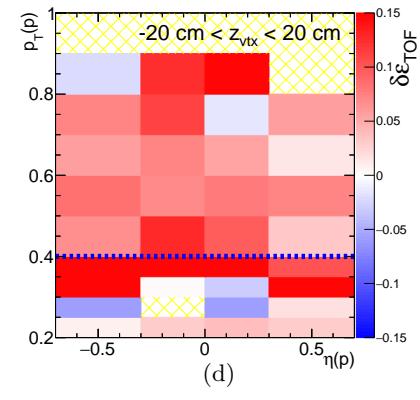
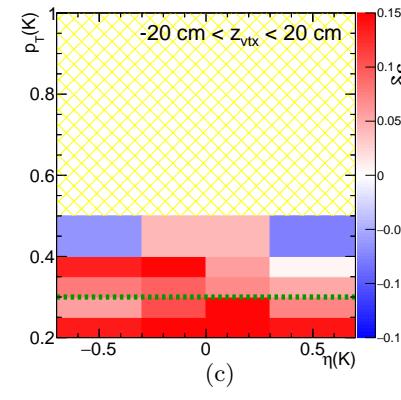
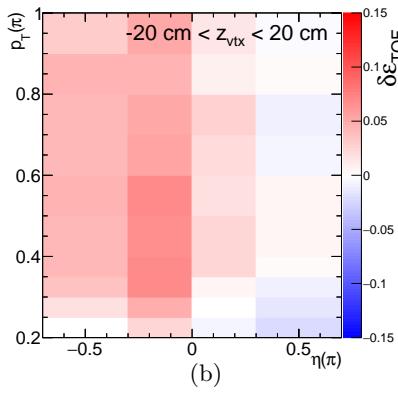
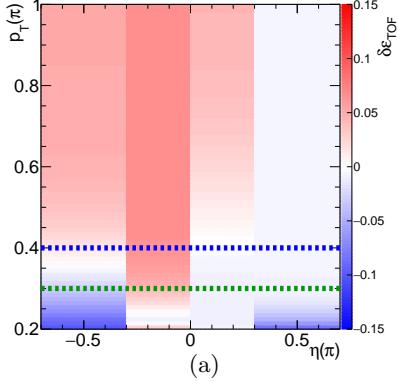


Figure 10.11: Temporary.

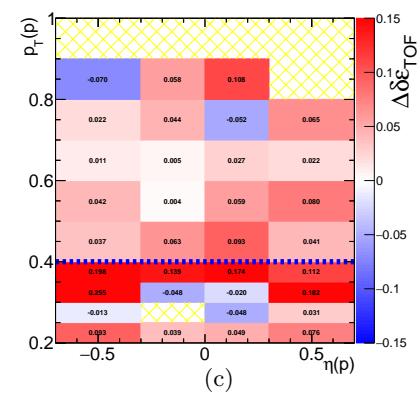
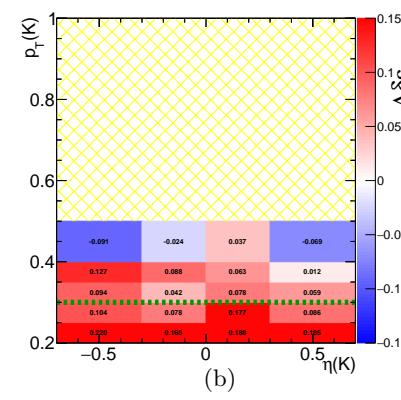
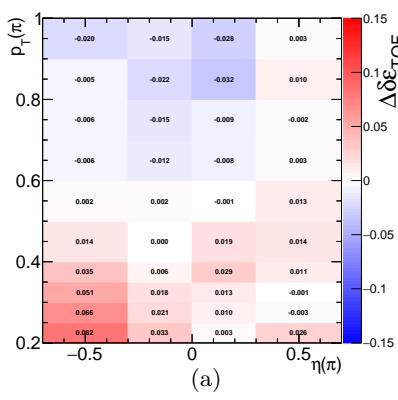
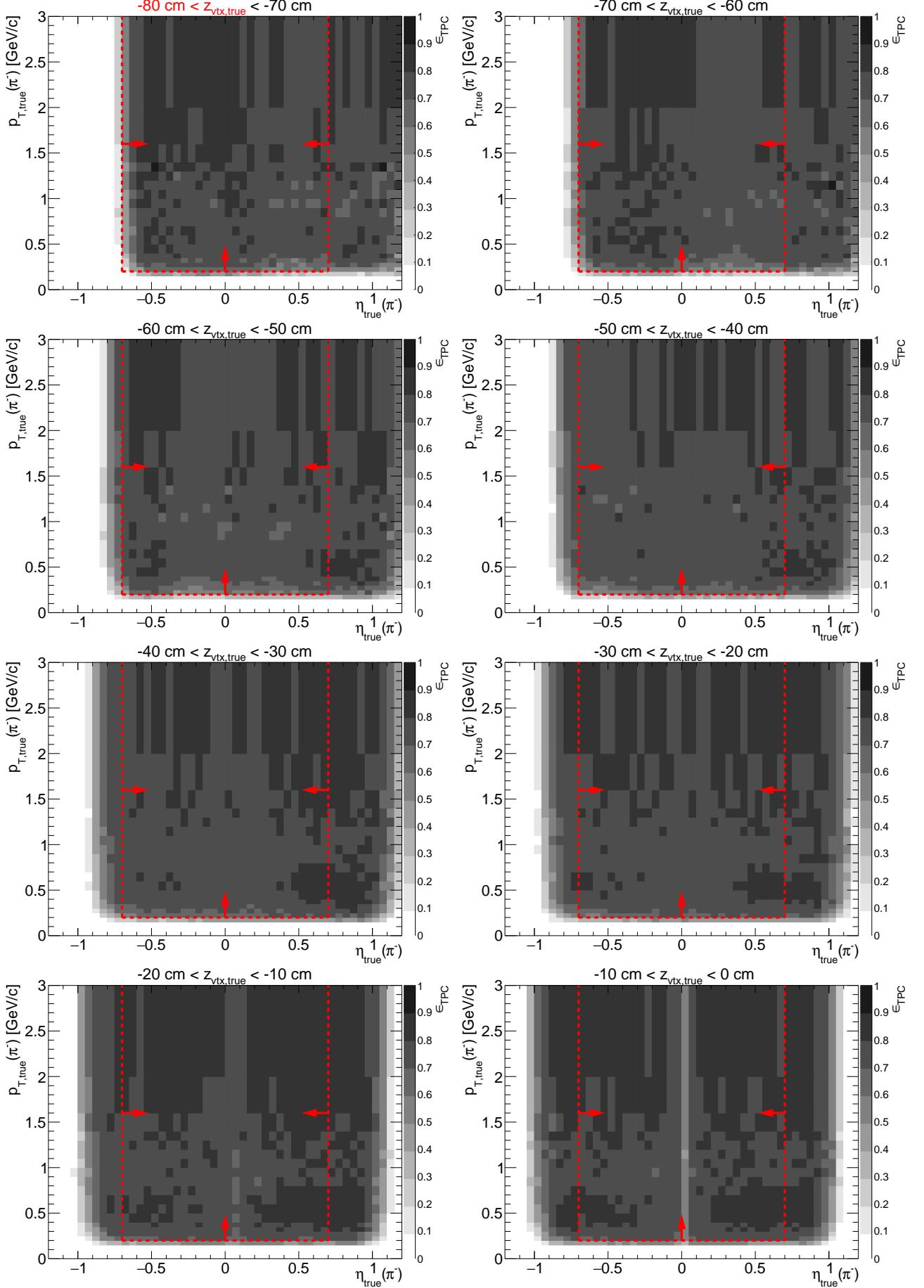


Figure 10.12: Temporary.

Appendix A

TPC track reconstruction efficiency

Figure A.1: TPC acceptance and reconstruction efficiency of π^- . Each plot represents the TPC efficiency ϵ_{TPC} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.



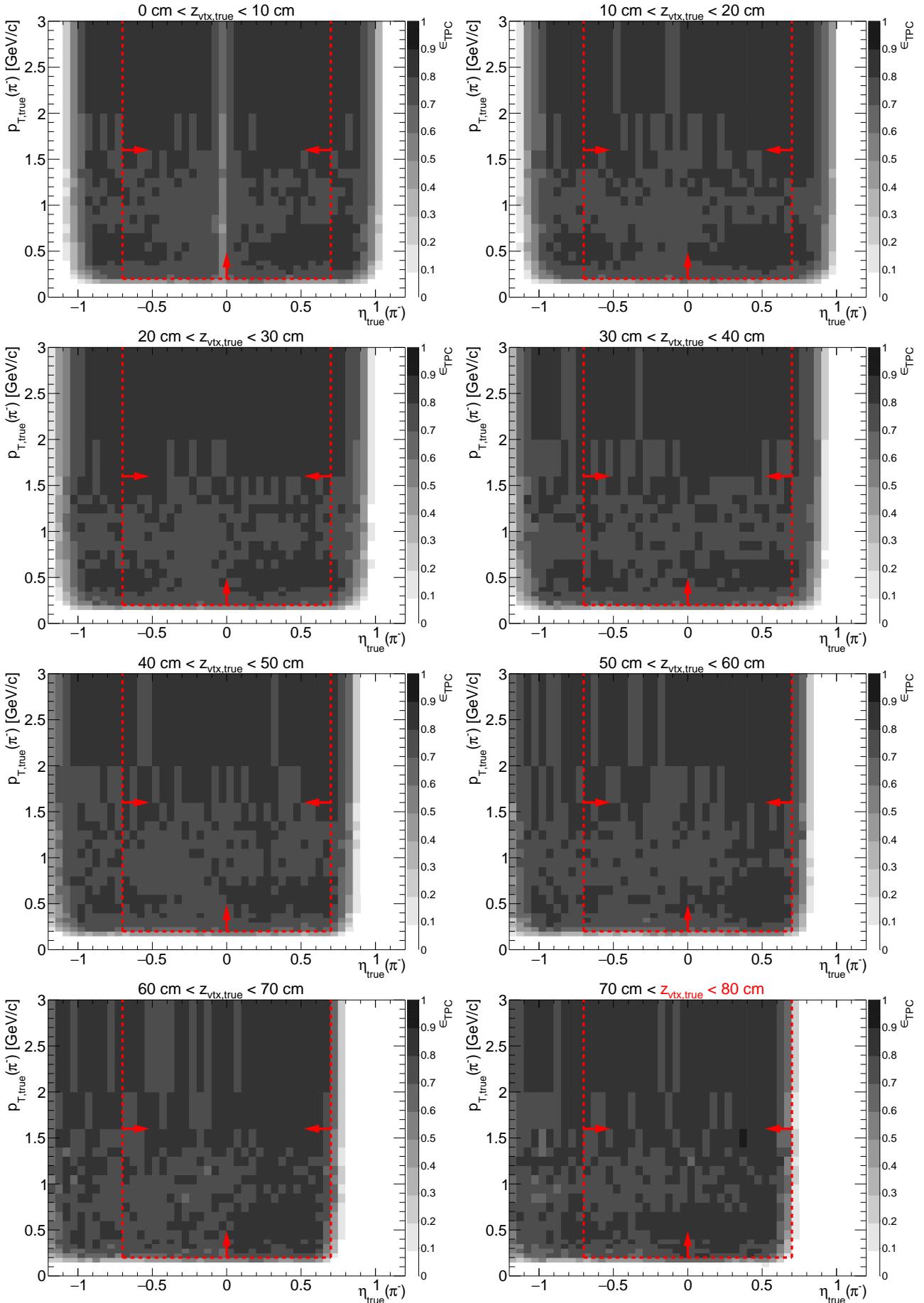
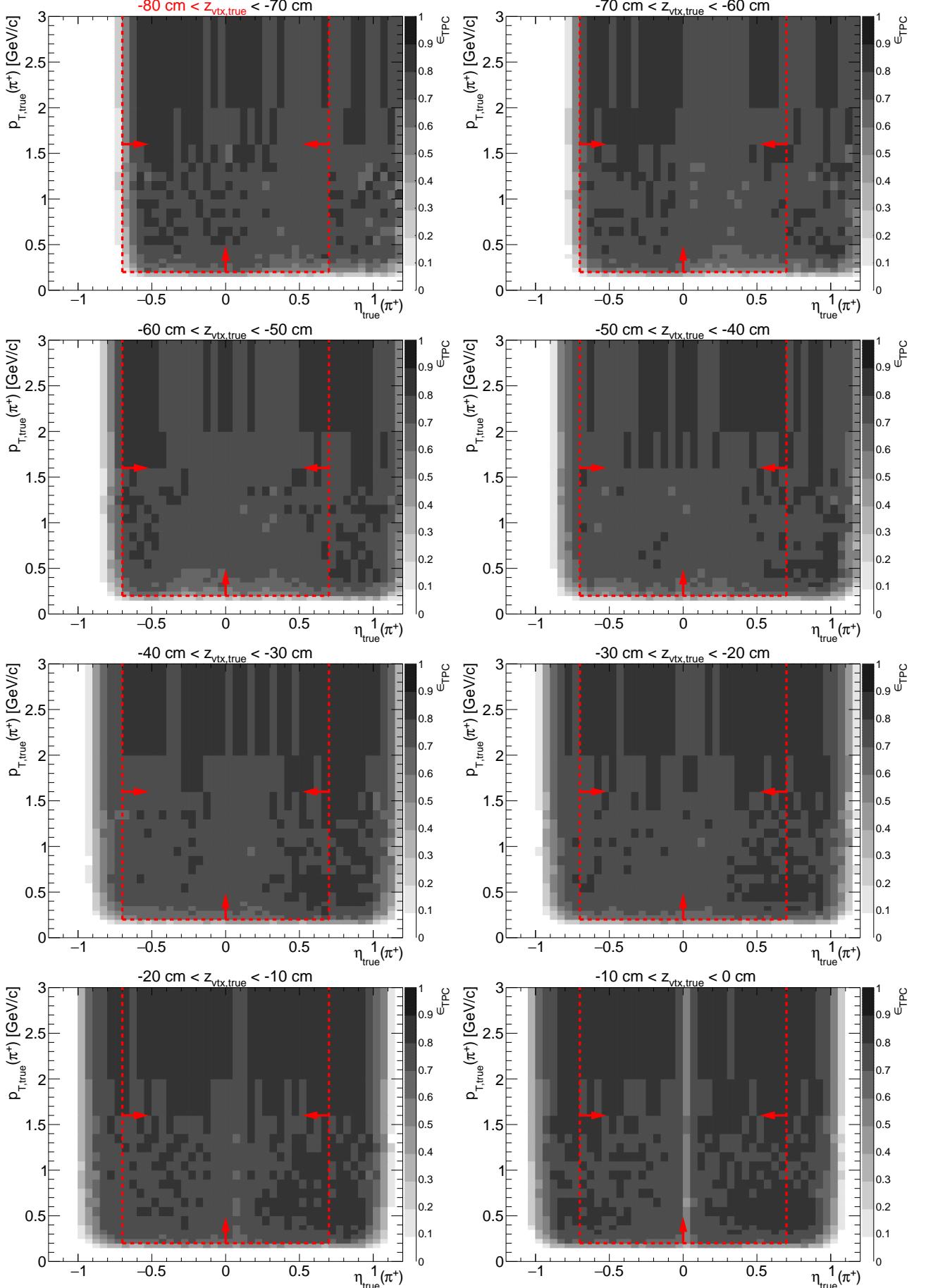


Figure A.2: TPC acceptance and reconstruction efficiency of π^+ . Each plot represents the TPC efficiency ϵ_{TPC} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.



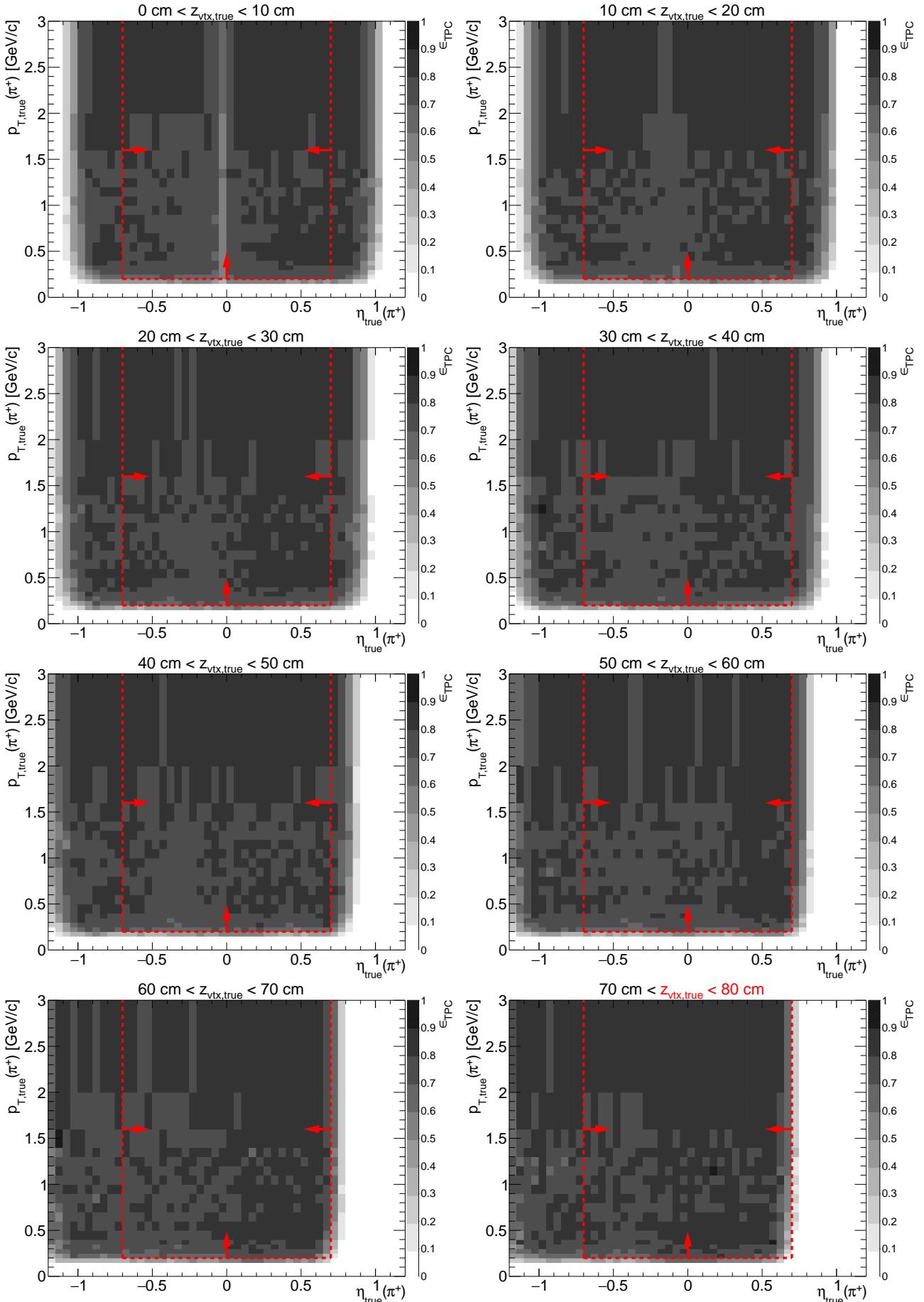
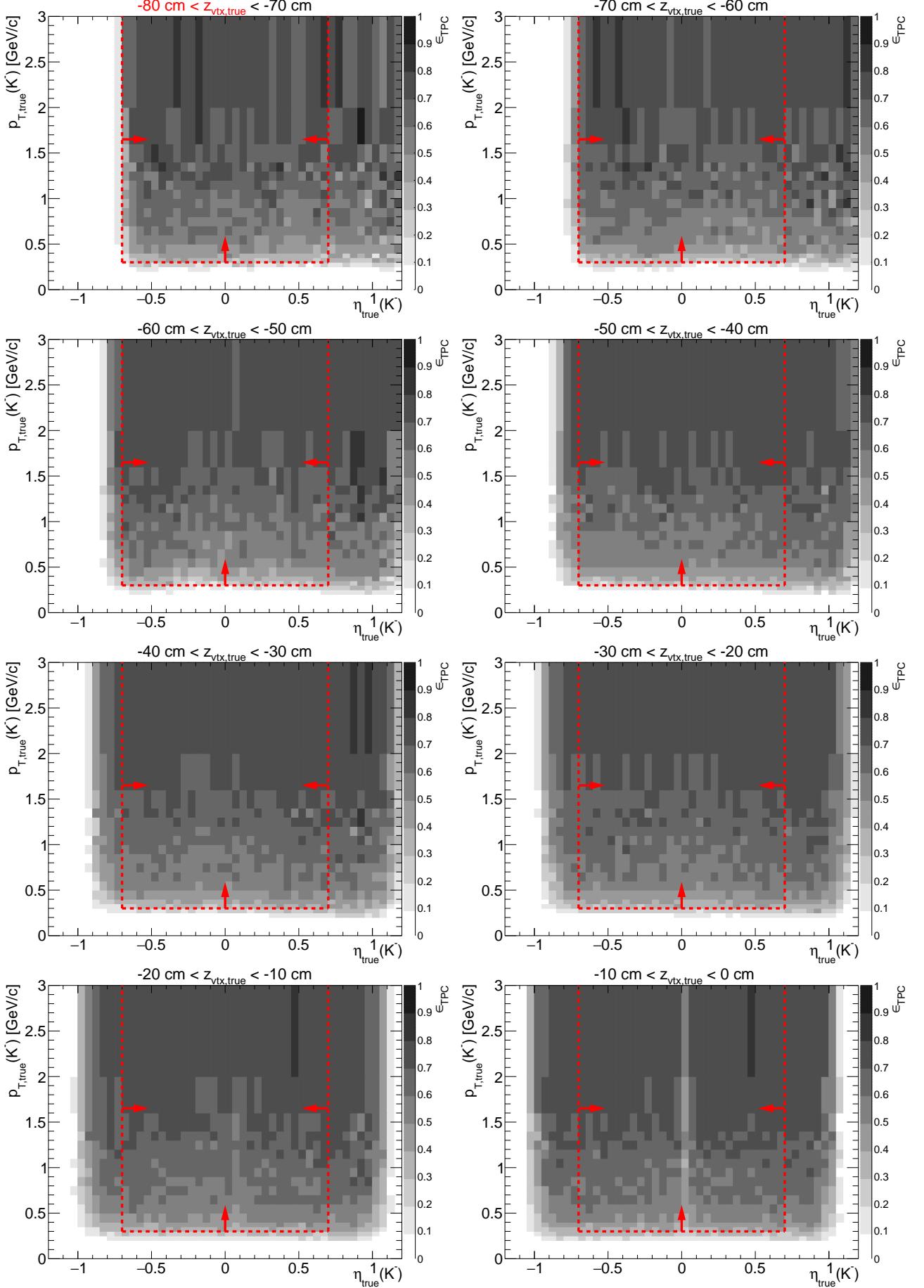


Figure A.3: TPC acceptance and reconstruction efficiency of K^- . Each plot represents the TPC efficiency ϵ_{TPC} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.



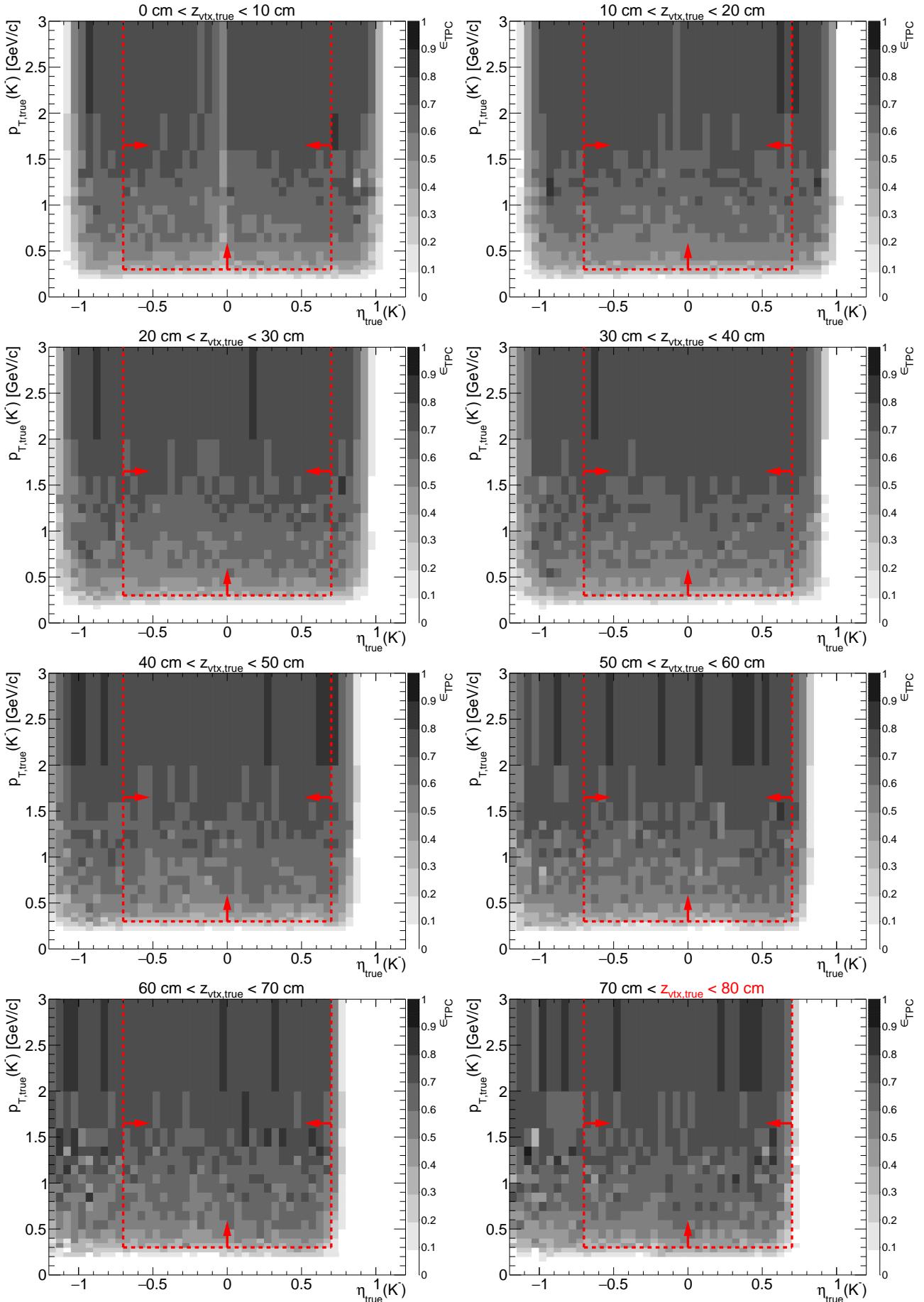
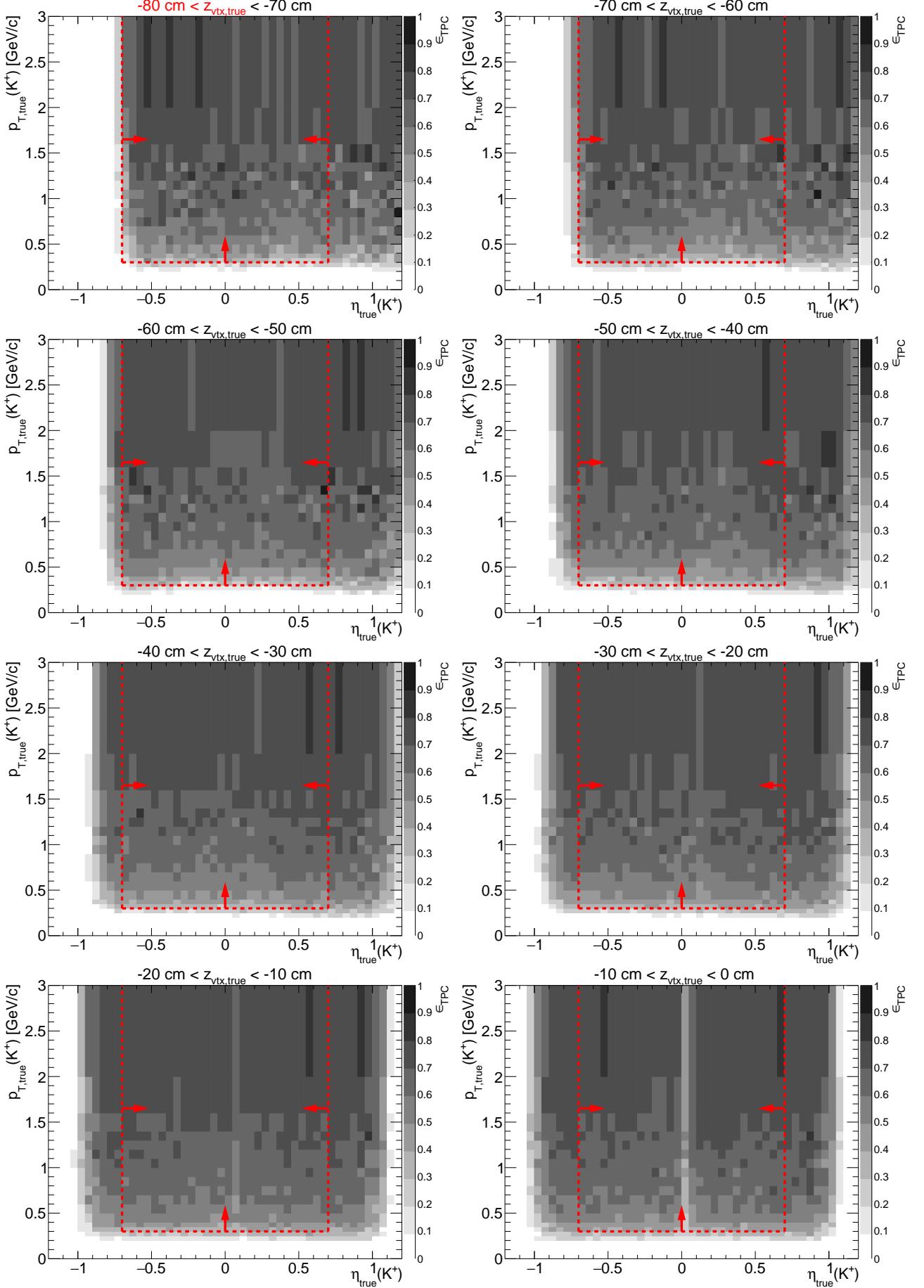


Figure A.4: TPC acceptance and reconstruction efficiency of K^+ . Each plot represents the TPC efficiency ϵ_{TPC} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.



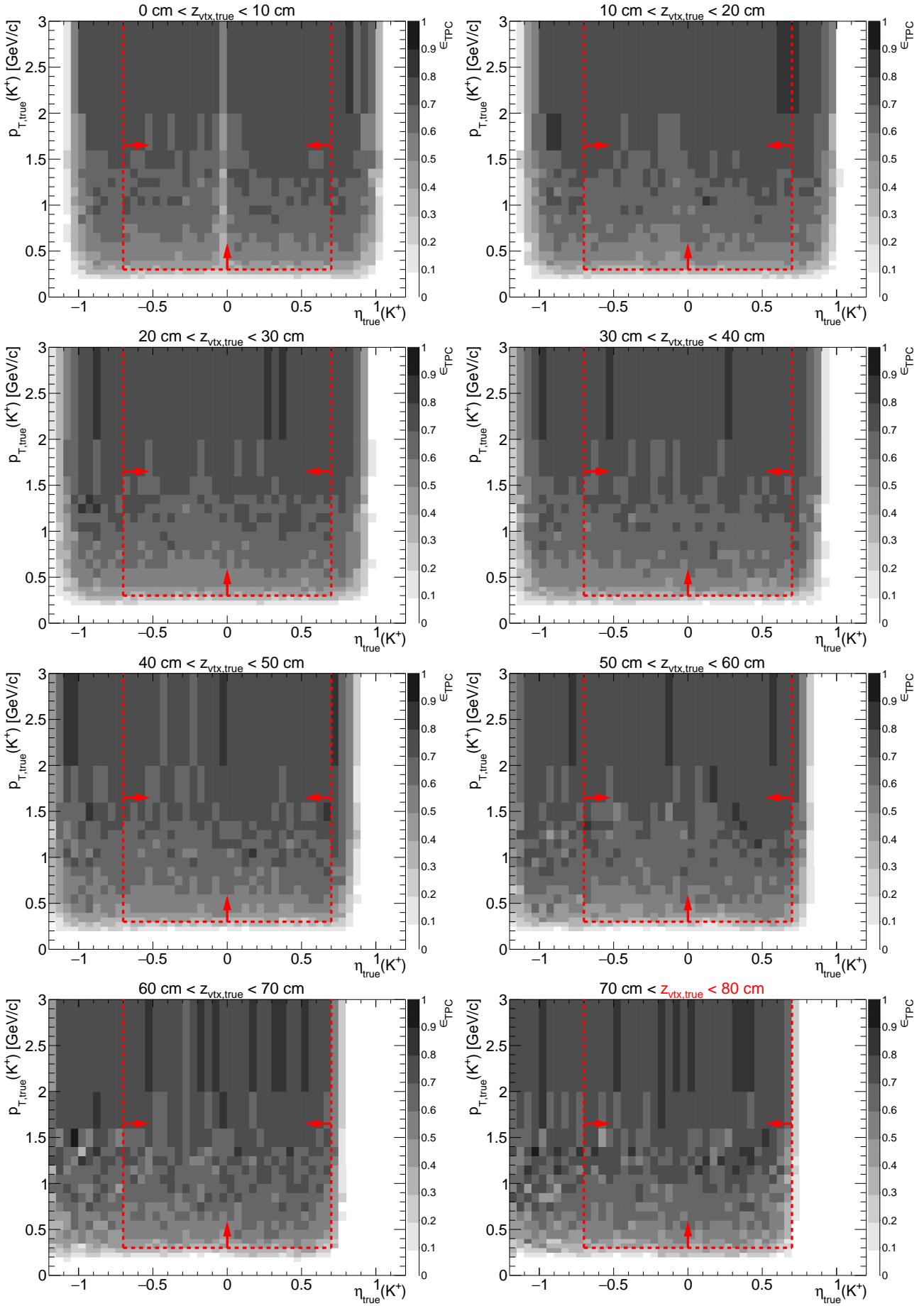
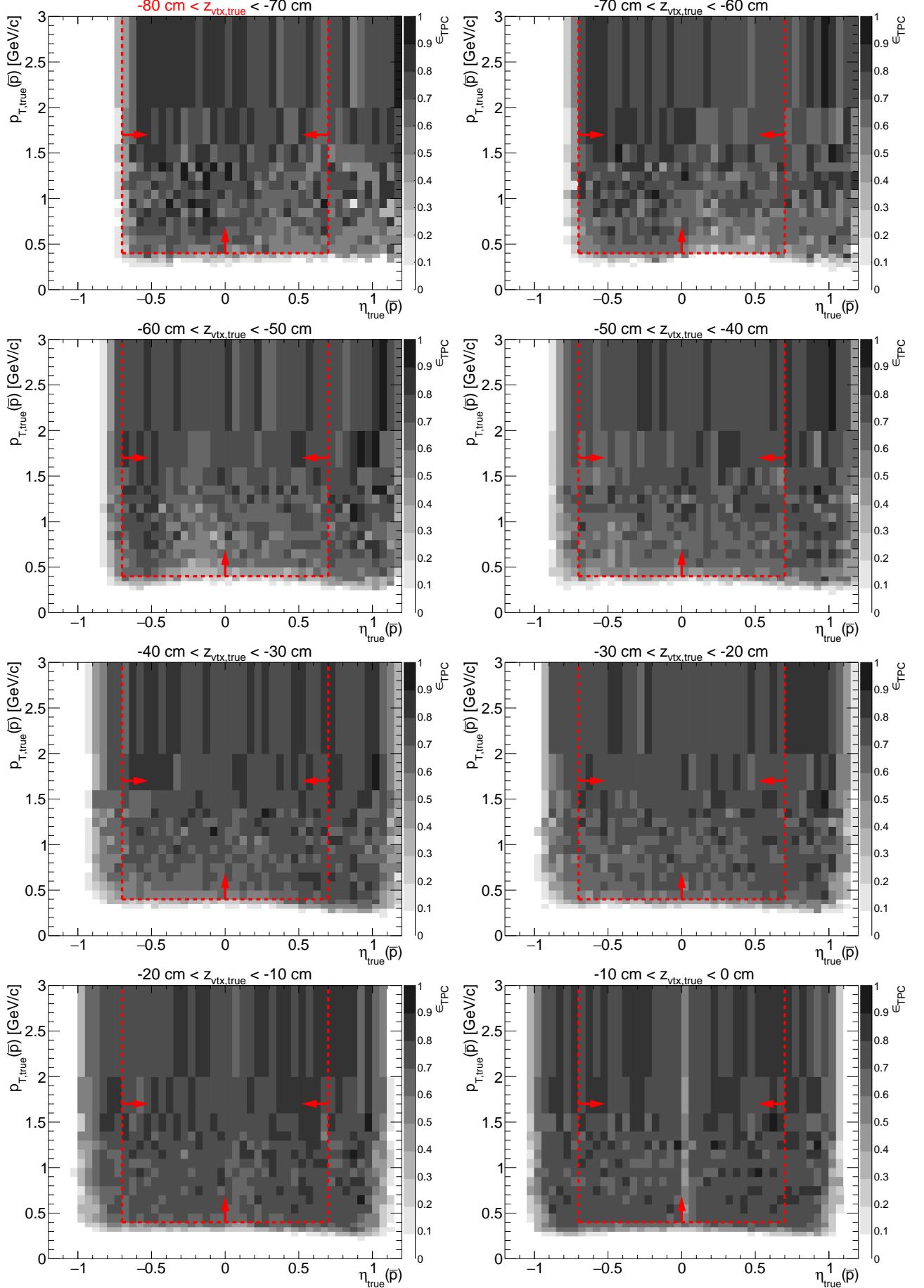


Figure A.5: TPC acceptance and reconstruction efficiency of \bar{p} . Each plot represents the TPC efficiency ϵ_{TPC} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.



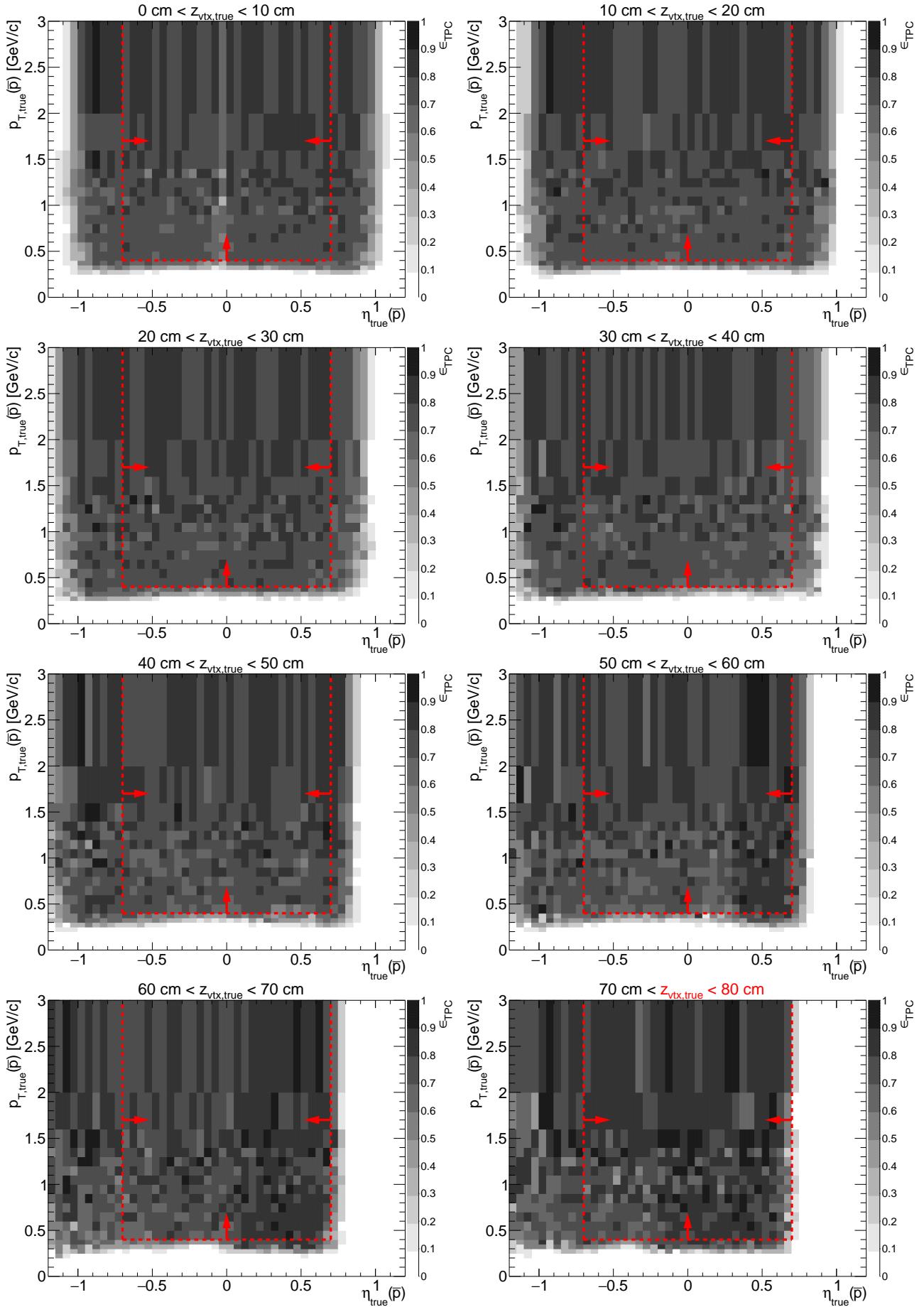
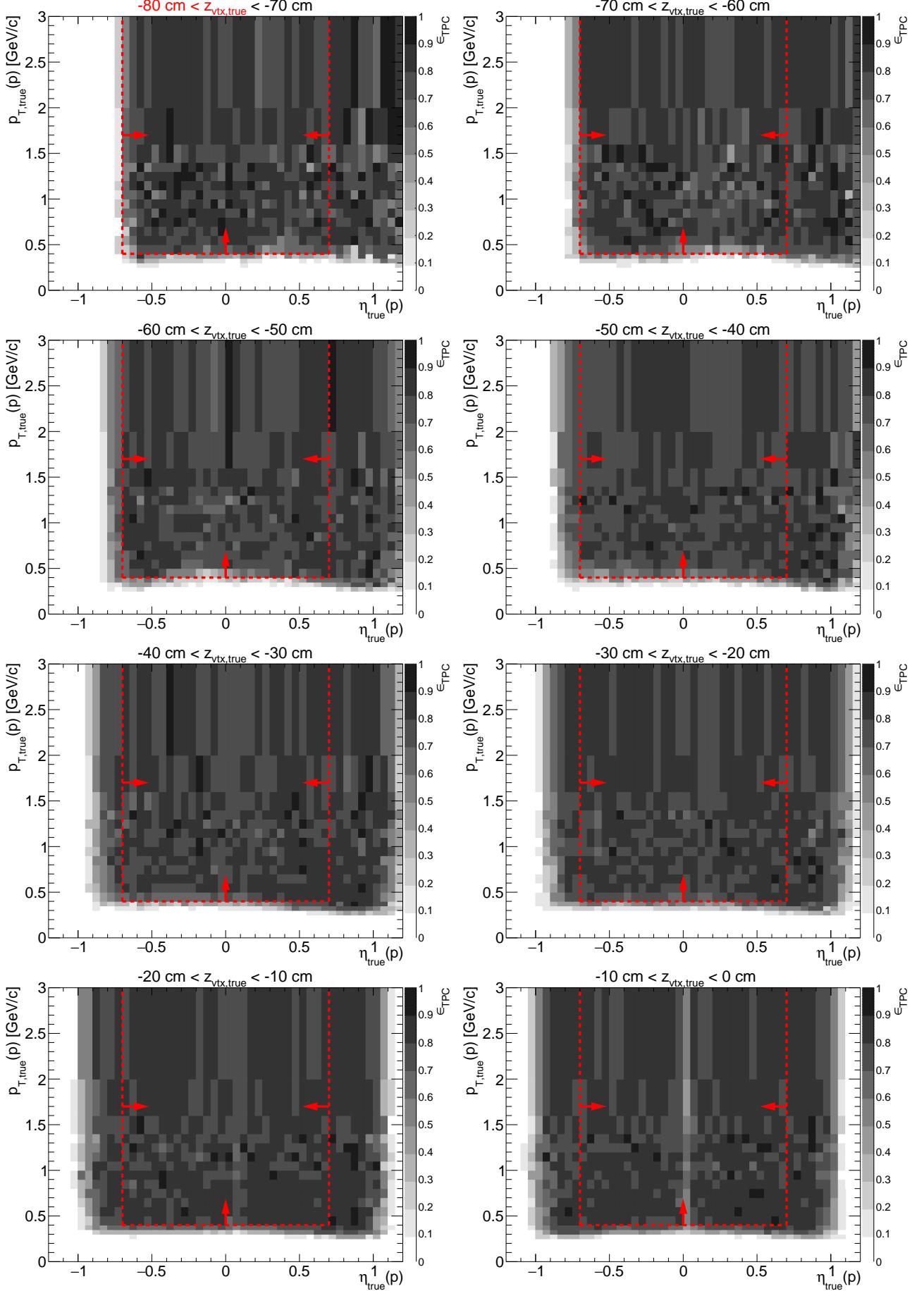
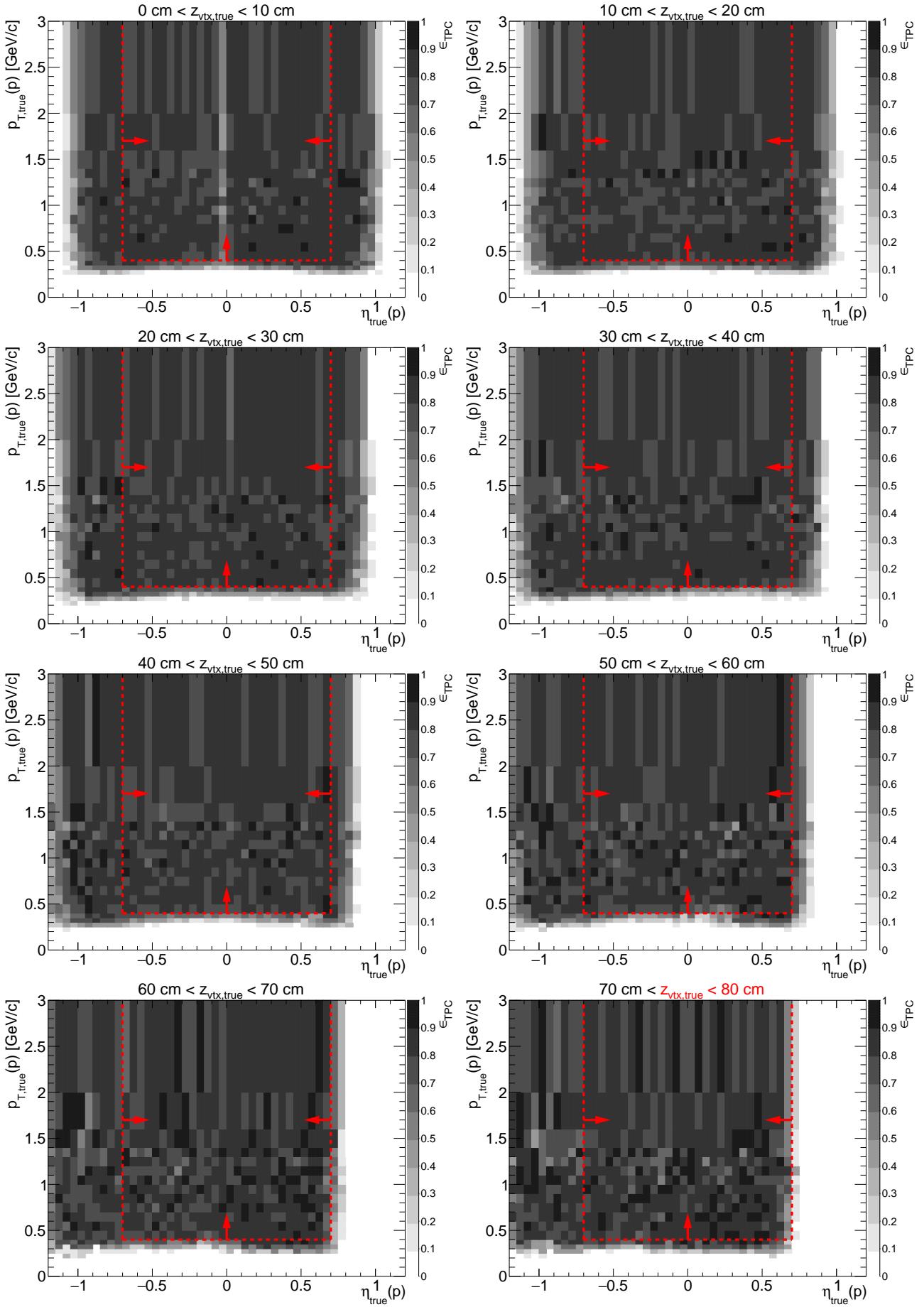


Figure A.6: TPC acceptance and reconstruction efficiency of p . Each plot represents the TPC efficiency ϵ_{TPC} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.





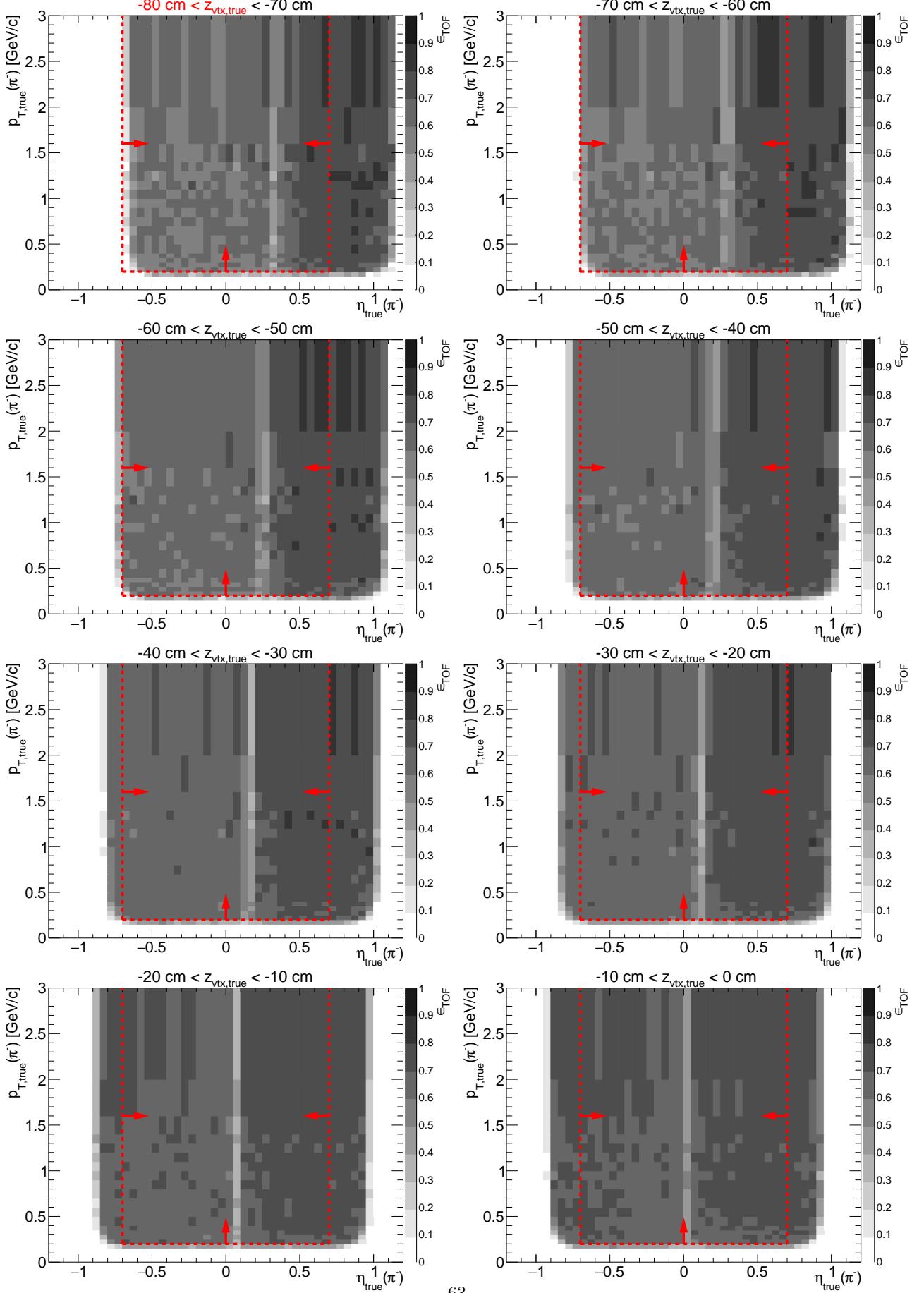
Appendix B

TOF hit reconstruction and matching efficiency

Tray No.	Module No.	RHIC fills
8	1 – 4, 29 – 32	18686 – 18953
23	1 – 4, 29 – 32	18686 – 18953
38	1 – 32	18686 – 18953
39	13	18686 – 18715, 18719 – 18795, 18797 – 18827, 18829 – 18854, 18856 – 18878, 18883 – 18892, 18895 – 18901, 18904 – 18924, 18926 – 18953
41	1 – 32	18686 – 18953
45	1 – 32	18877 – 18909
46	1 – 32	18853 – 18909
93	1 – 4, 29 – 32	18686 – 18953
102	1 – 32	18686 – 18953
108	1 – 4, 29 – 32	18686 – 18953

Table B.1: Dead TOF modules masked in the MC. Table was filled with modules which were not matched with TPC tracks in the data.

Figure B.1: TOF acceptance, reconstruction and matching efficiency of π^- . Each plot represents the TOF efficiency ϵ_{TOF} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.



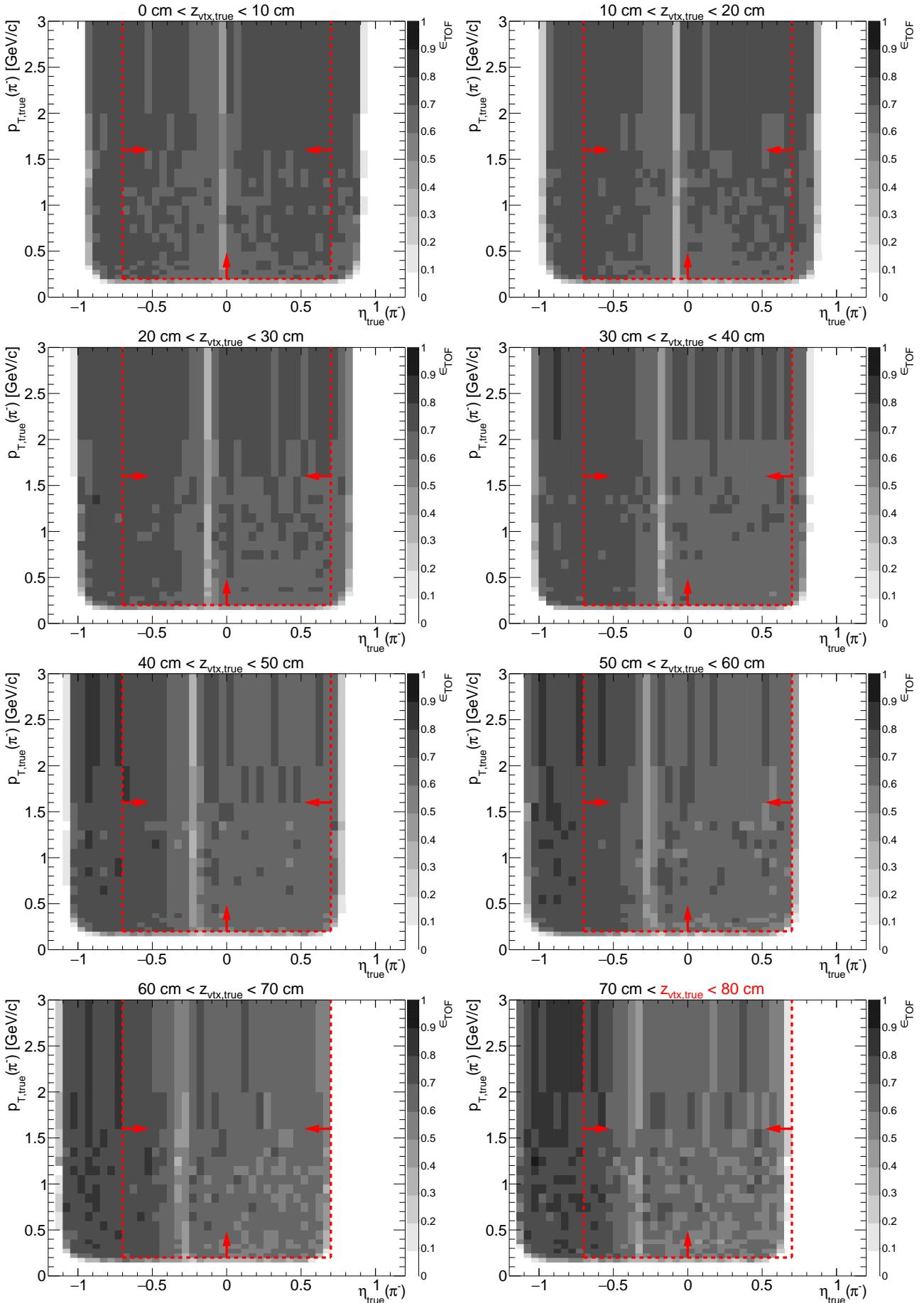
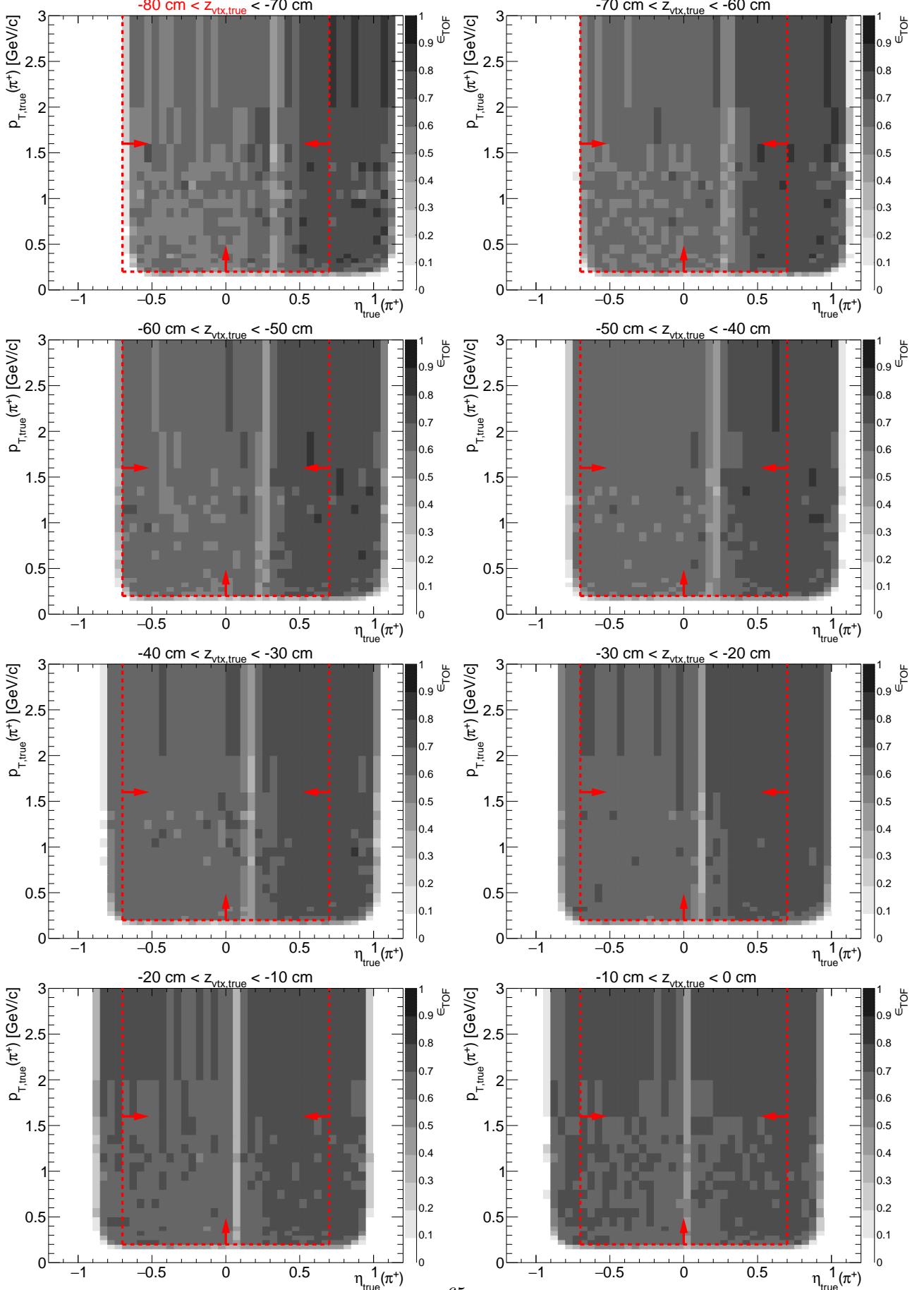


Figure B.2: TOF acceptance, reconstruction and matching efficiency of π^+ . Each plot represents the TOF efficiency ϵ_{TOF} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.



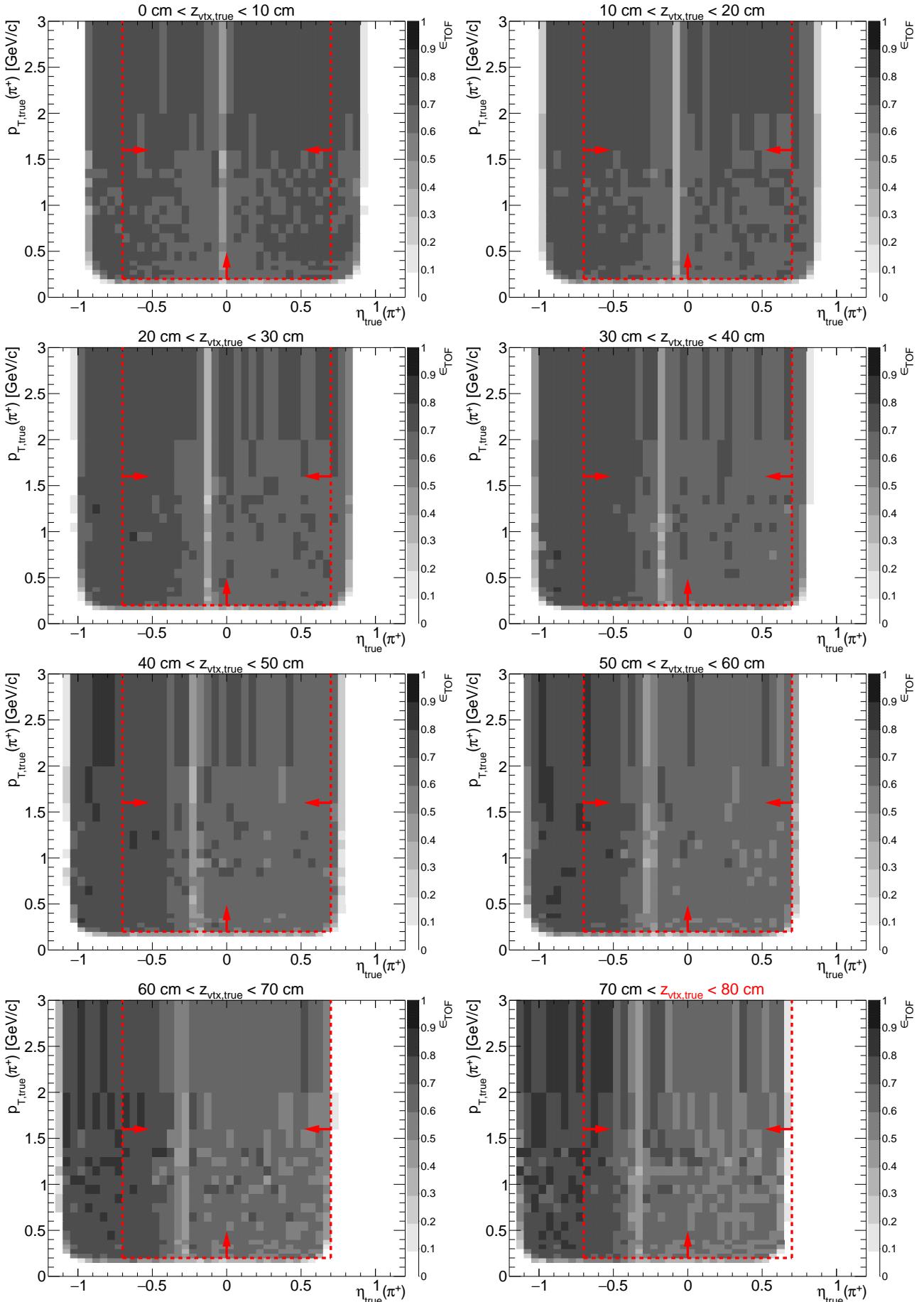
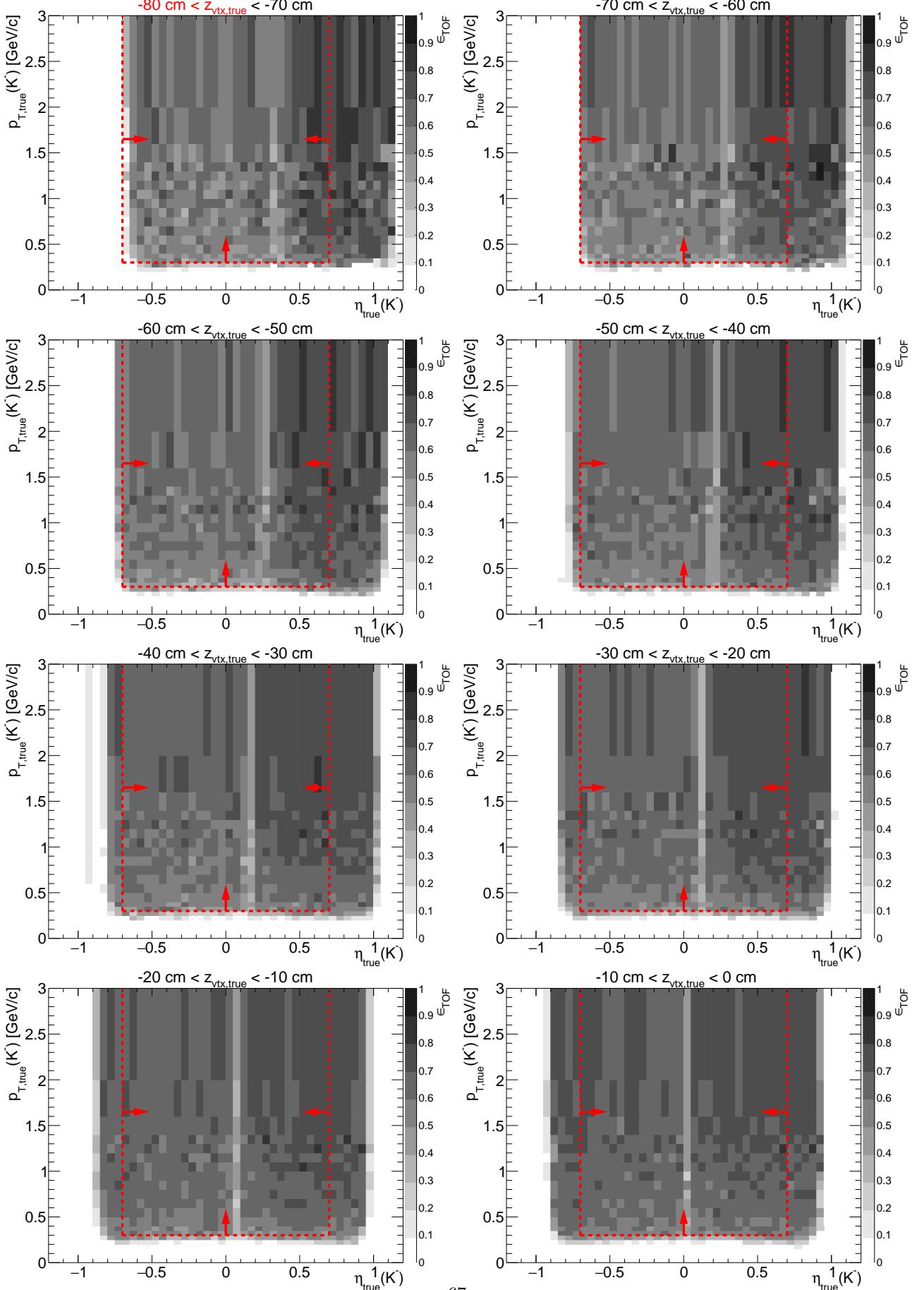


Figure B.3: TOF acceptance, reconstruction and matching efficiency of K^- . Each plot represents the TOF efficiency ϵ_{TOF} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.



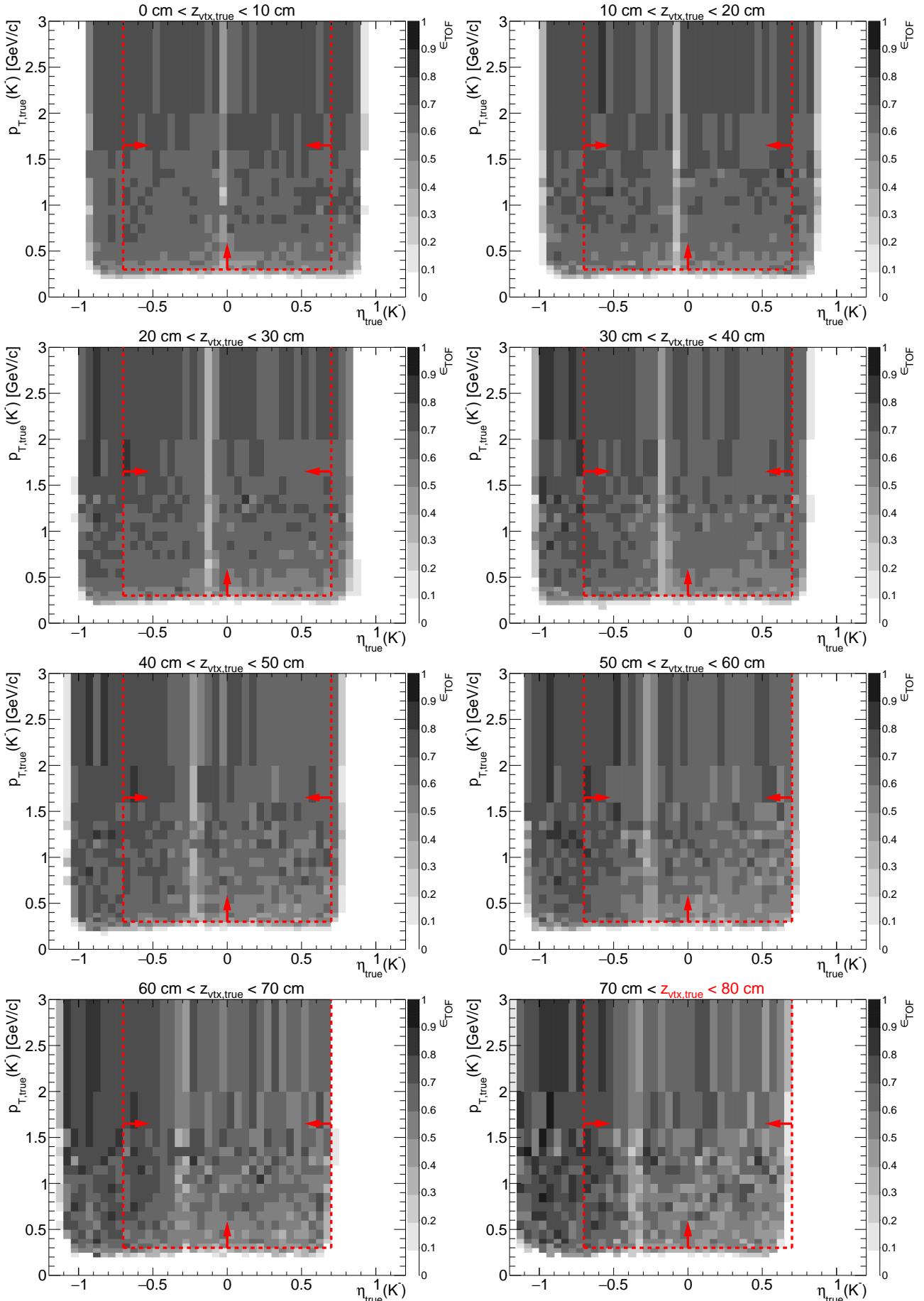
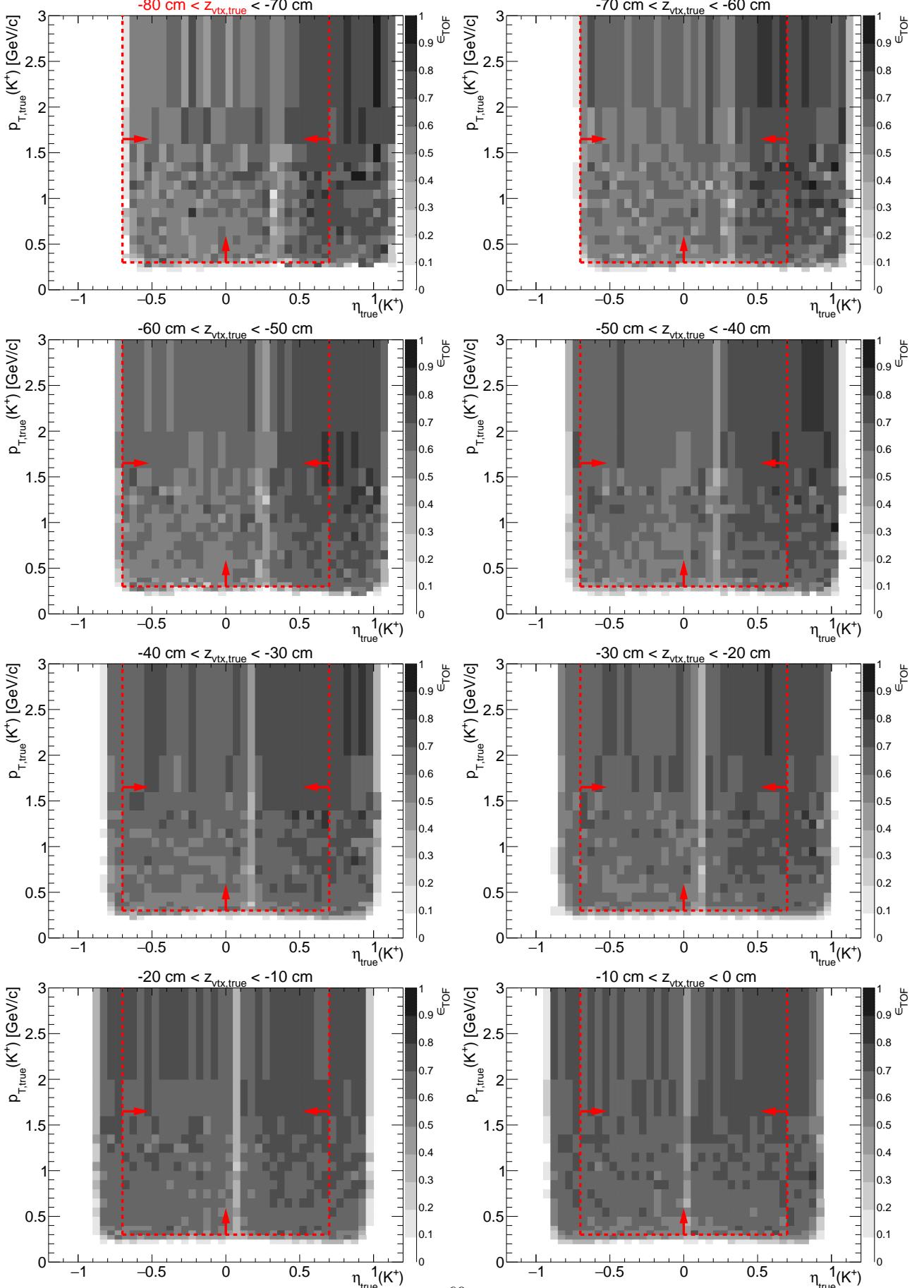


Figure B.4: TOF acceptance, reconstruction and matching efficiency of K^+ . Each plot represents the TOF efficiency ϵ_{TOF} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.



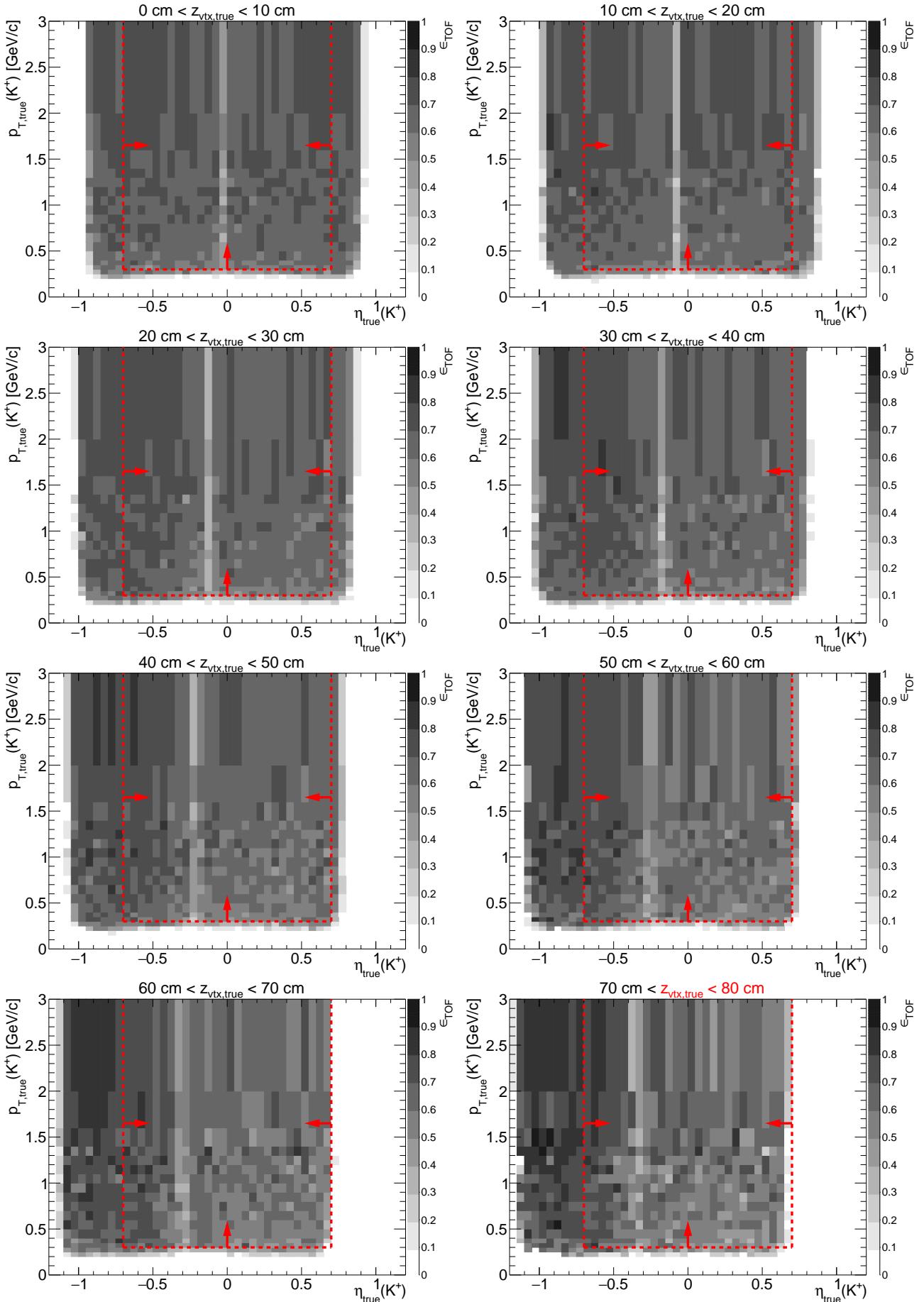
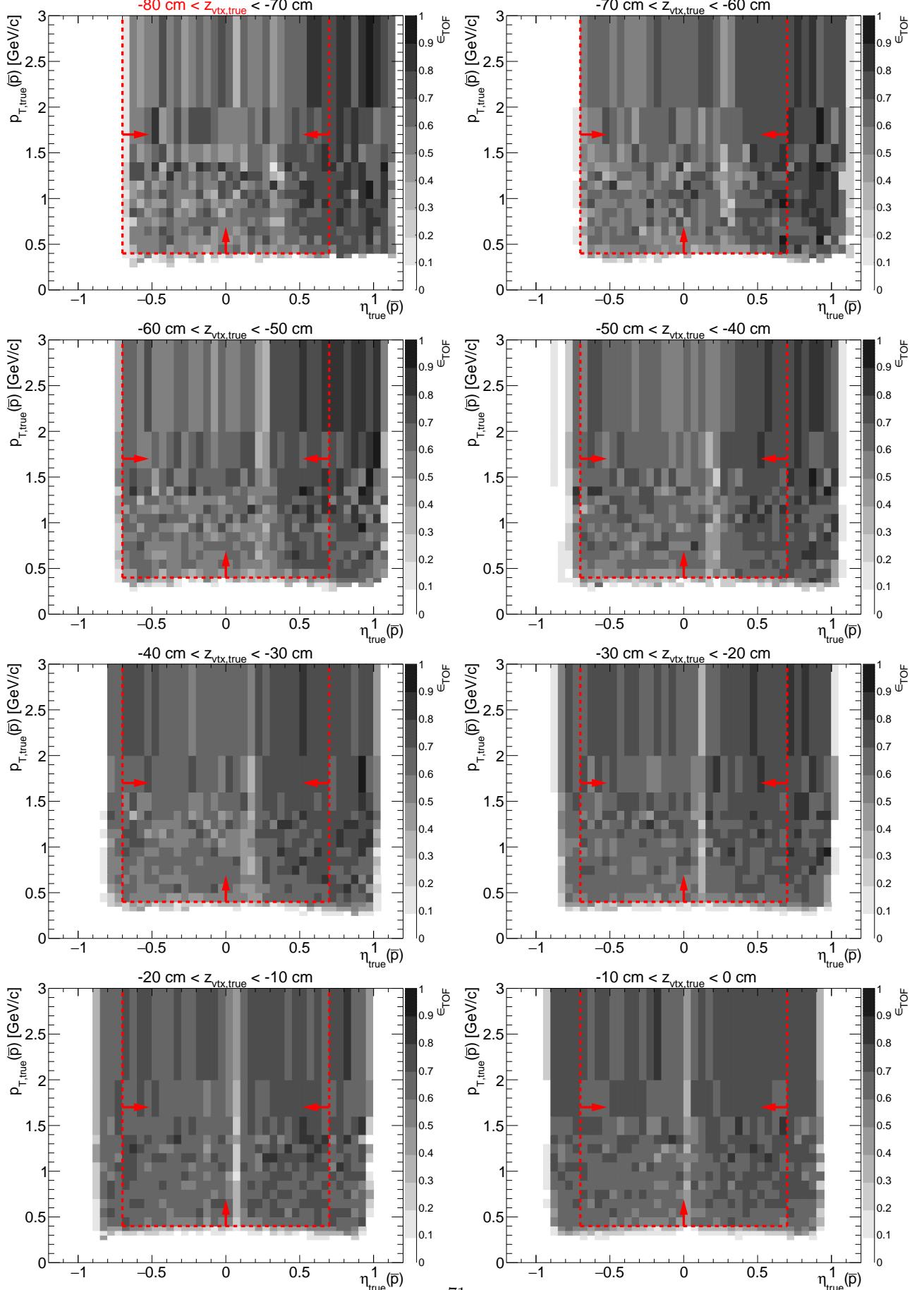


Figure B.5: TOF acceptance, reconstruction and matching efficiency of \bar{p} . Each plot represents the TOF efficiency ϵ_{TOF} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.



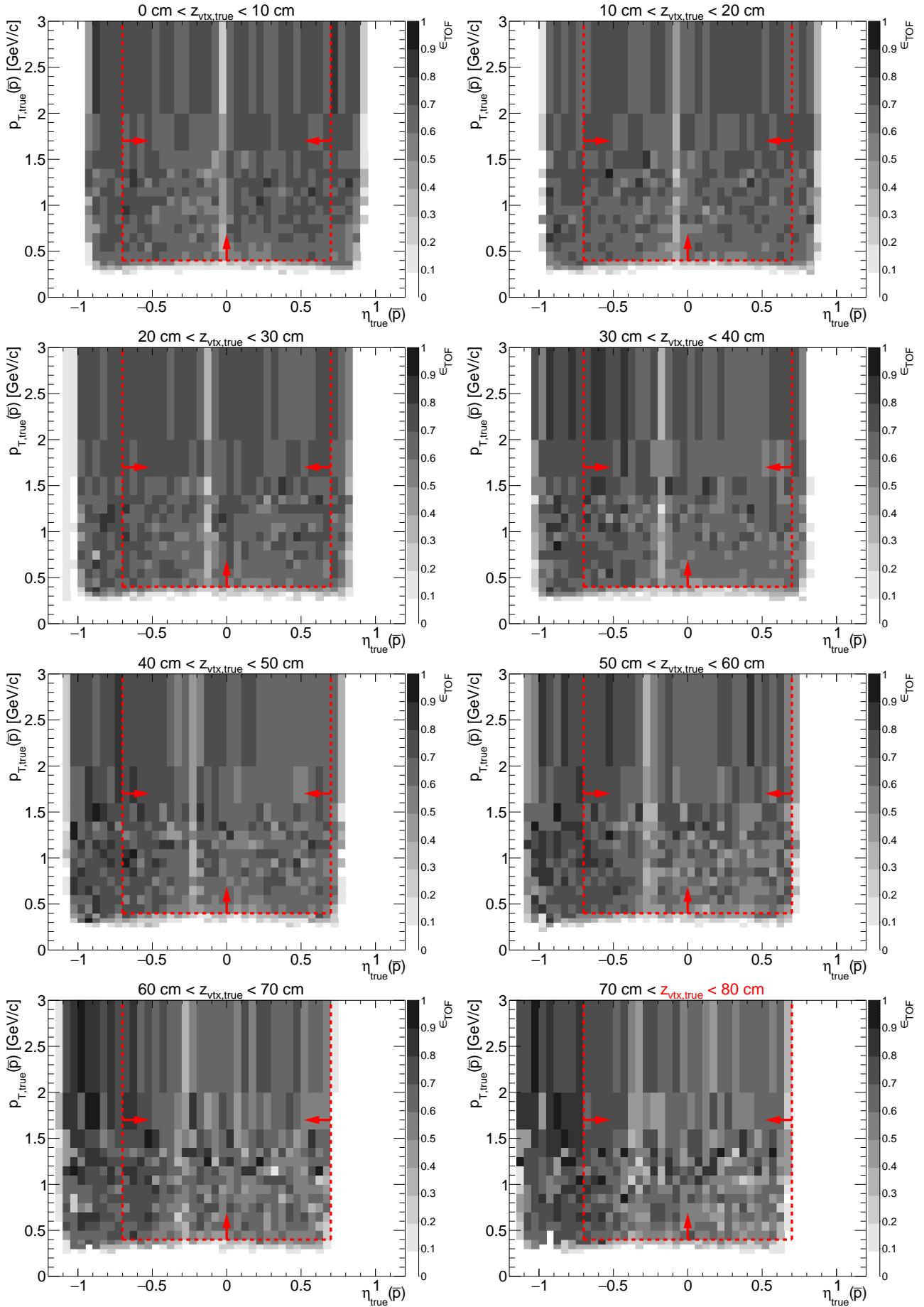
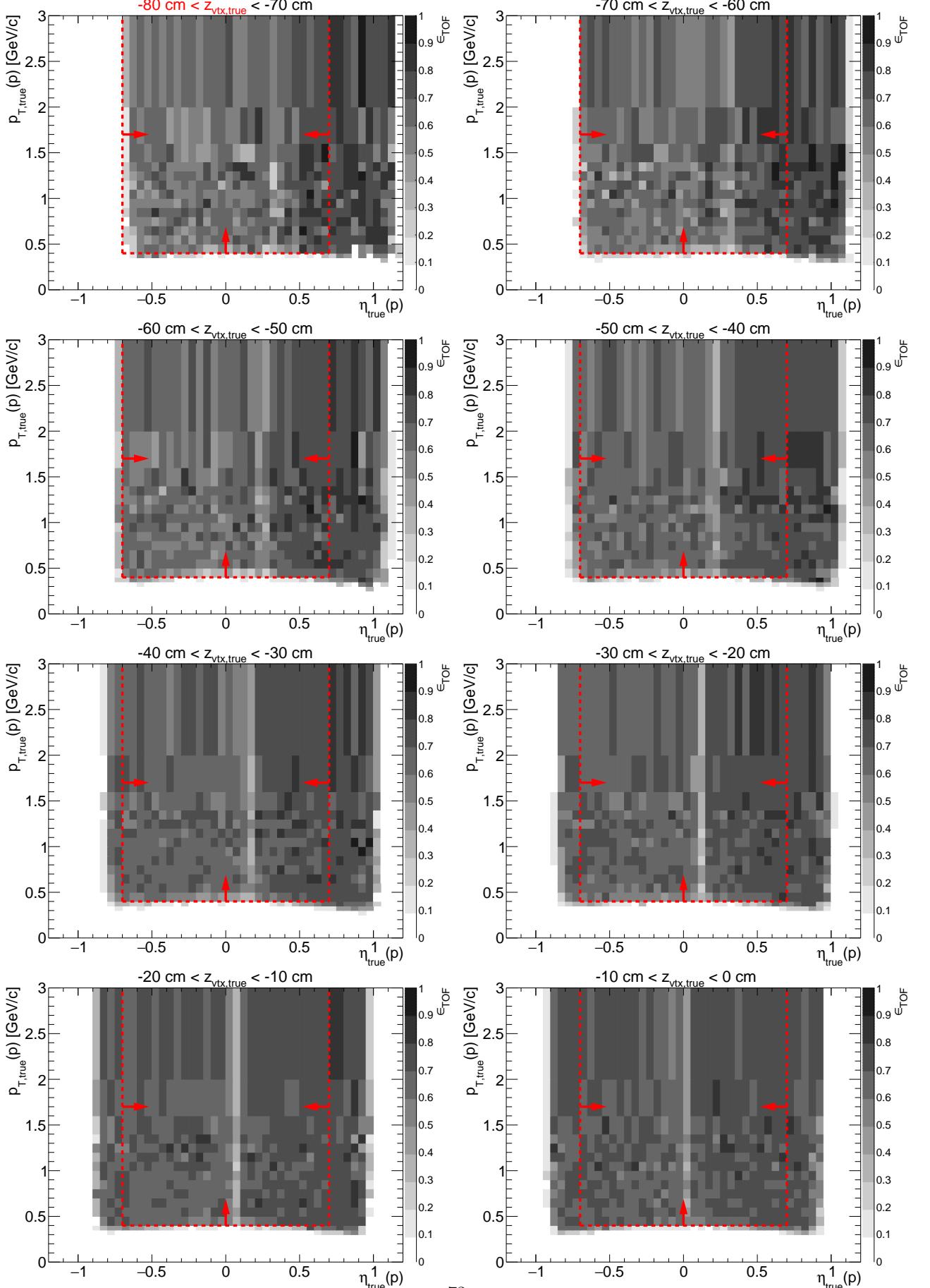
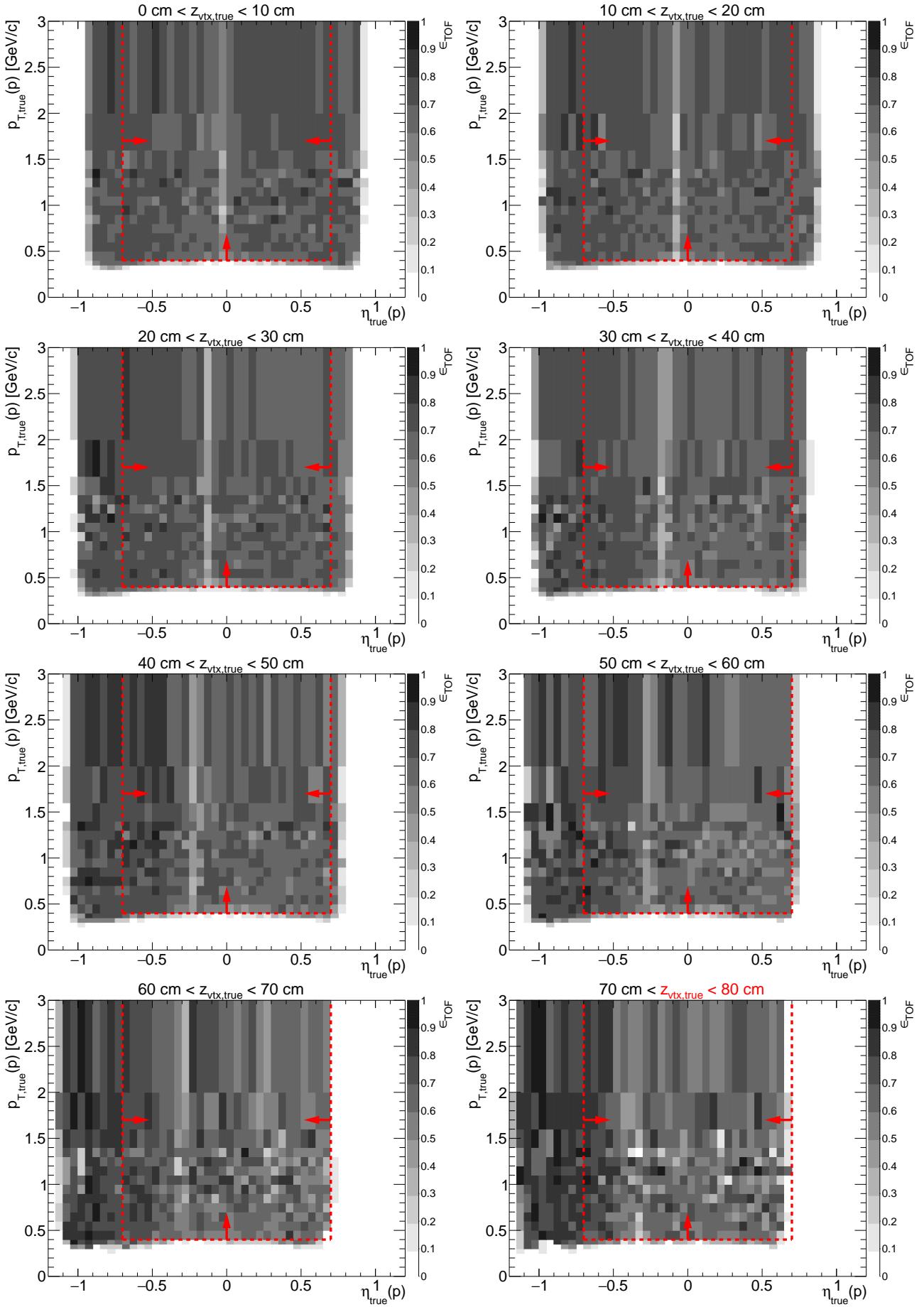


Figure B.6: TOF acceptance, reconstruction and matching efficiency of p . Each plot represents the TOF efficiency ϵ_{TOF} (z -axis) as a function of true particle pseudorapidity η (x -axis) and transverse momentum p_T (y -axis) in single z -vertex bin whose range is given at the top. Red lines and arrows indicate region accepted in analyses.

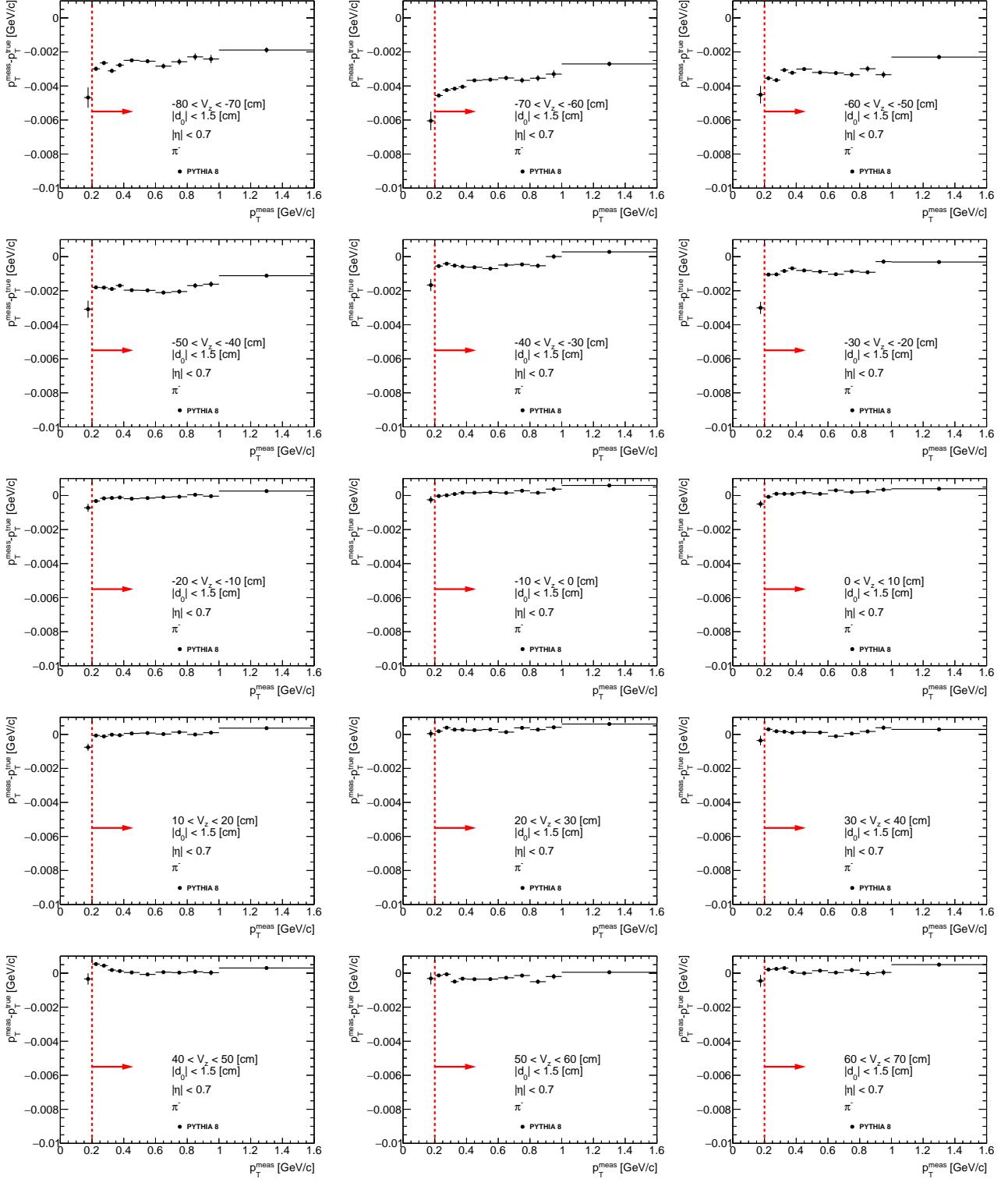




Appendix C

Energy Loss Correction

Figure C.1: Energy loss correction $p_T^{meas} - p_T^{true}$ for π^- as a function of reconstructed transverse momentum p_T^{meas} ($|\eta| < 0.7$) in single z -vertex bin whose range is given on each plot. Red lines and arrows indicate region accepted in analyses. One can notice an offset of about $3 - 4$ MeV for negative z -vertex. It is a known issue with STAR simulation where HFT support material is badly described for negative $z < -30$ cm.



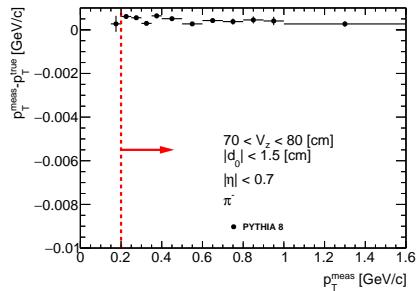


Figure C.2: Energy loss correction $p_T^{meas} - p_T^{true}$ for π^+ as a function of reconstructed transverse momentum p_T^{meas} ($|\eta| < 0.7$) in single z -vertex bin whose range is given on each plot. Red lines and arrows indicate region accepted in analyses. One can notice an offset of about 3 – 4 MeV for negative z -vertex. It is a known issue with STAR simulation where HFT support material is badly described for negative $z < -30$ cm.

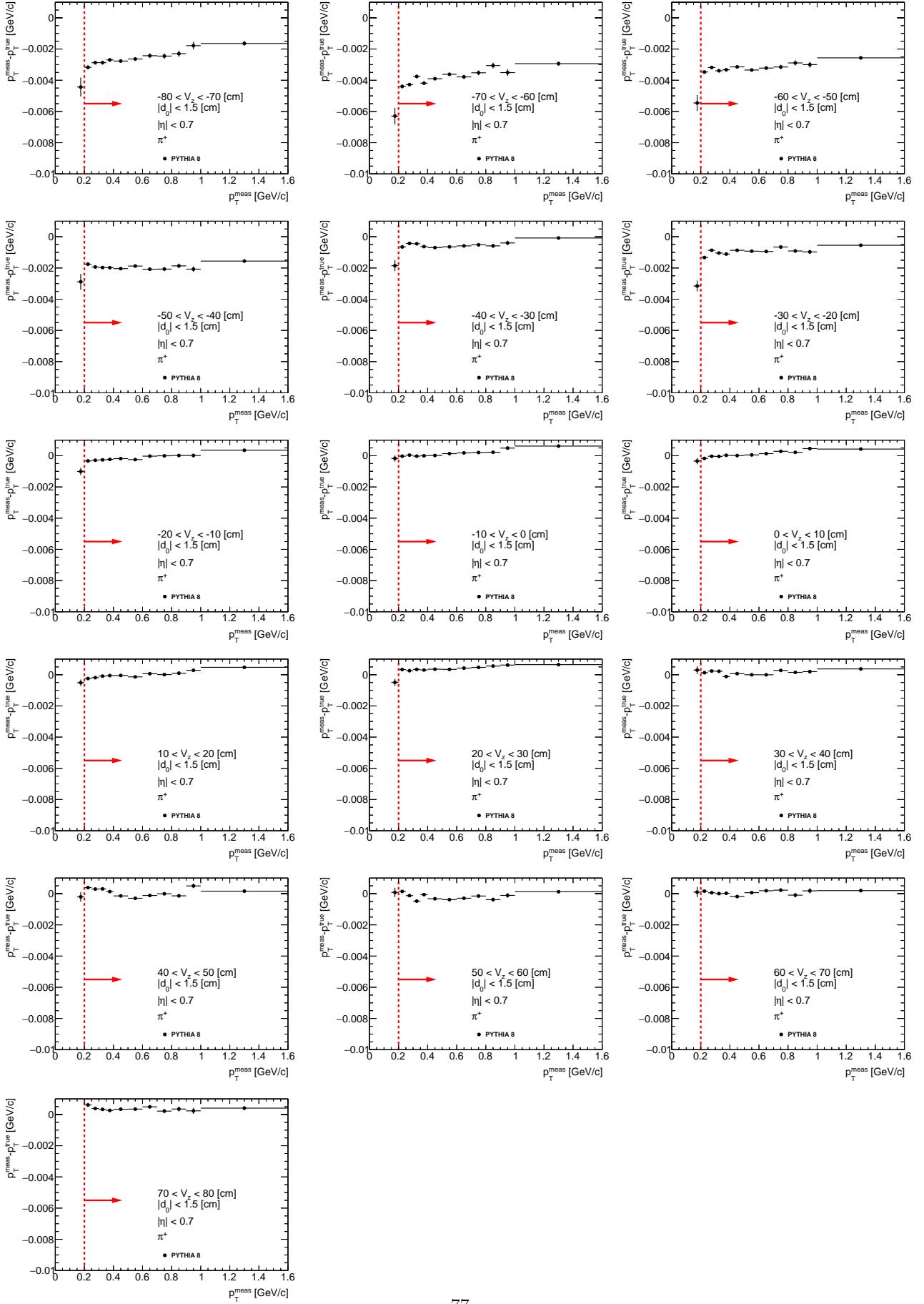


Figure C.3: Energy loss correction $p_T^{meas} - p_T^{true}$ for K^- as a function of reconstructed transverse momentum p_T^{meas} ($|\eta| < 0.7$) in single z -vertex bin whose range is given on each plot. Red lines and arrows indicate region accepted in analyses.

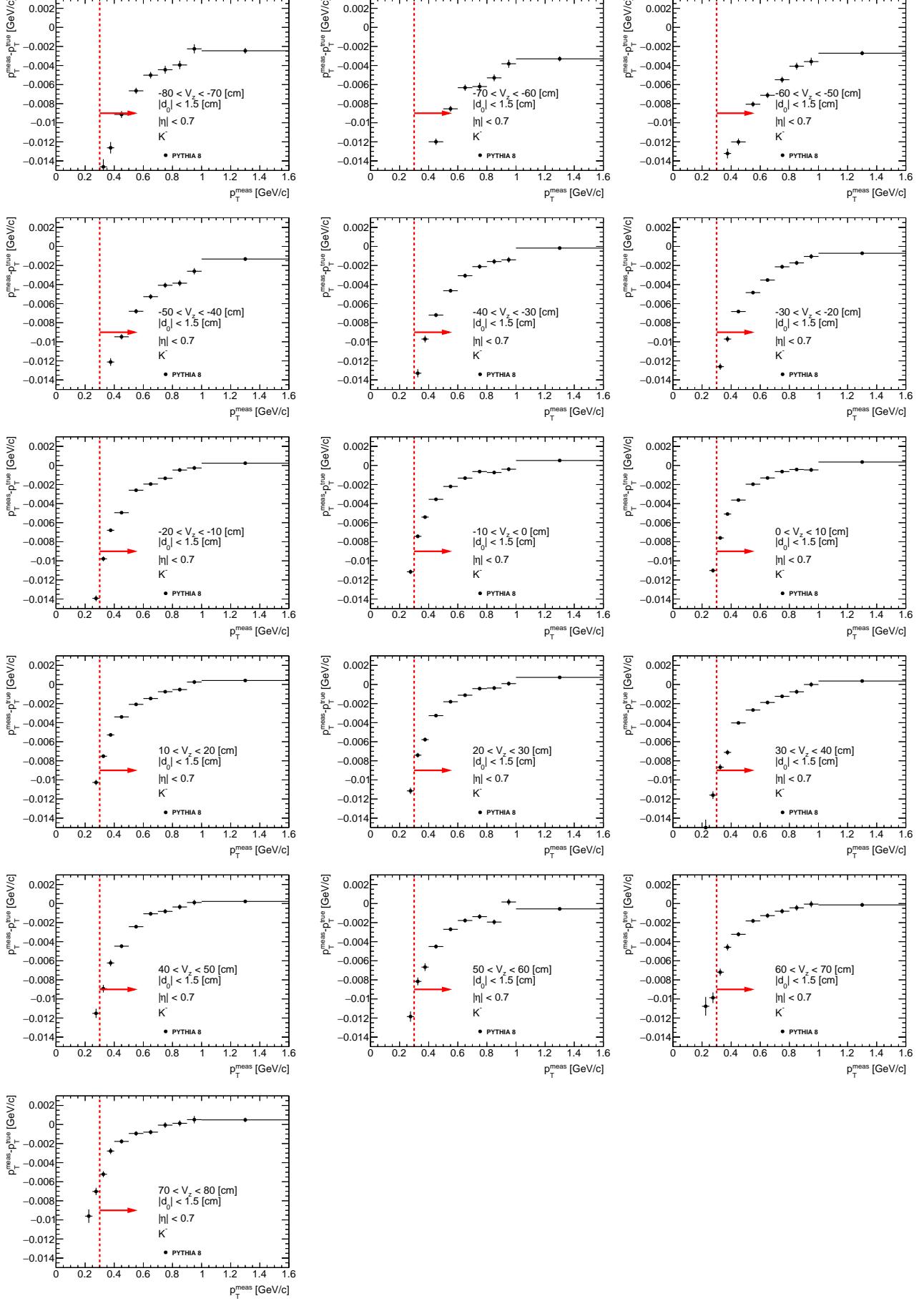


Figure C.4: Energy loss correction $p_T^{meas} - p_T^{true}$ for K^+ as a function of reconstructed transverse momentum p_T^{meas} ($|\eta| < 0.7$) in single z -vertex bin whose range is given on each plot. Red lines and arrows indicate region accepted in analyses.

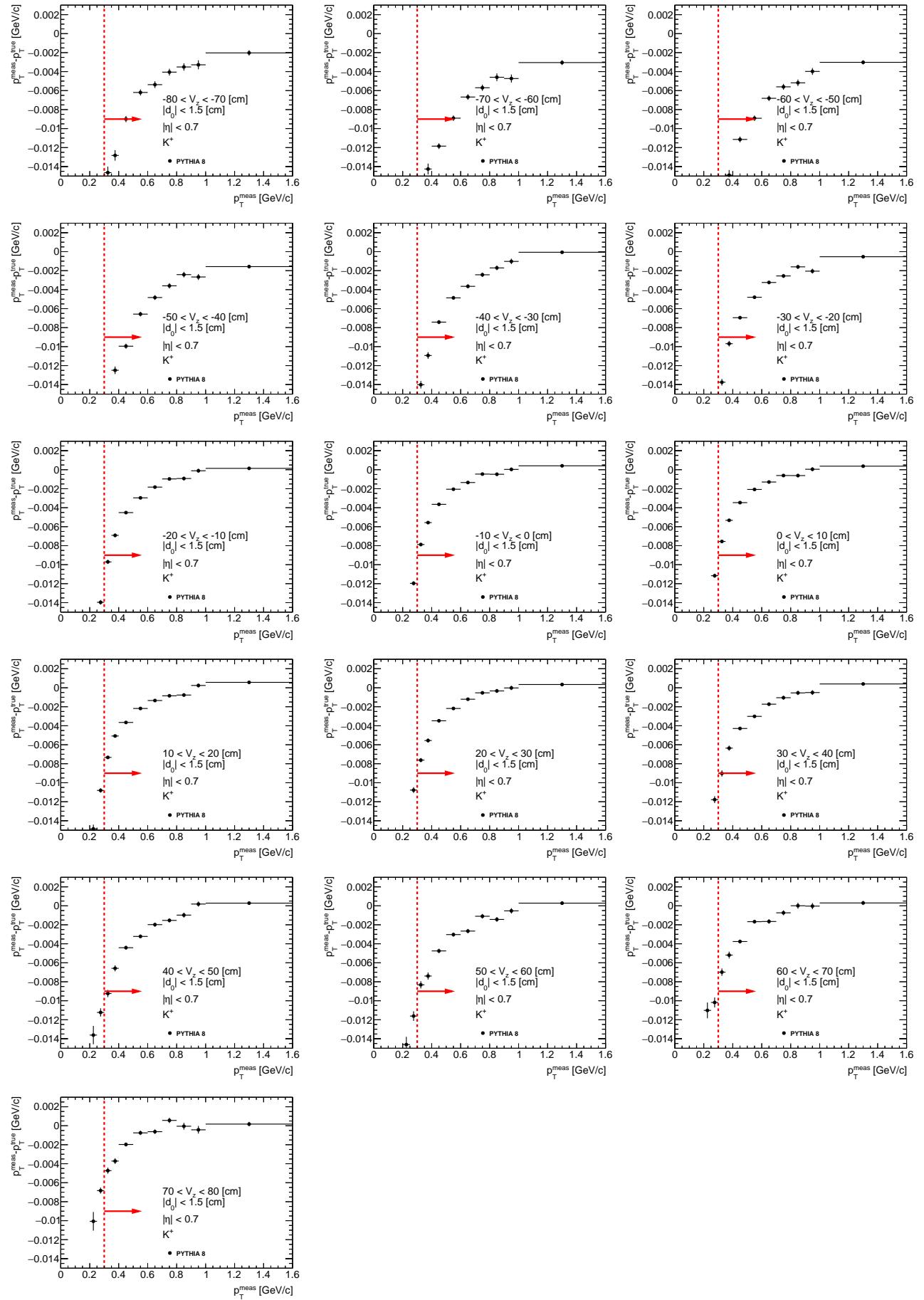


Figure C.5: Energy loss correction $p_T^{\text{meas}} - p_T^{\text{true}}$ for \bar{p} as a function of reconstructed transverse momentum p_T^{meas} ($|\eta| < 0.7$) in single z -vertex bin whose range is given on each plot. Red lines and arrows indicate region accepted in analyses.

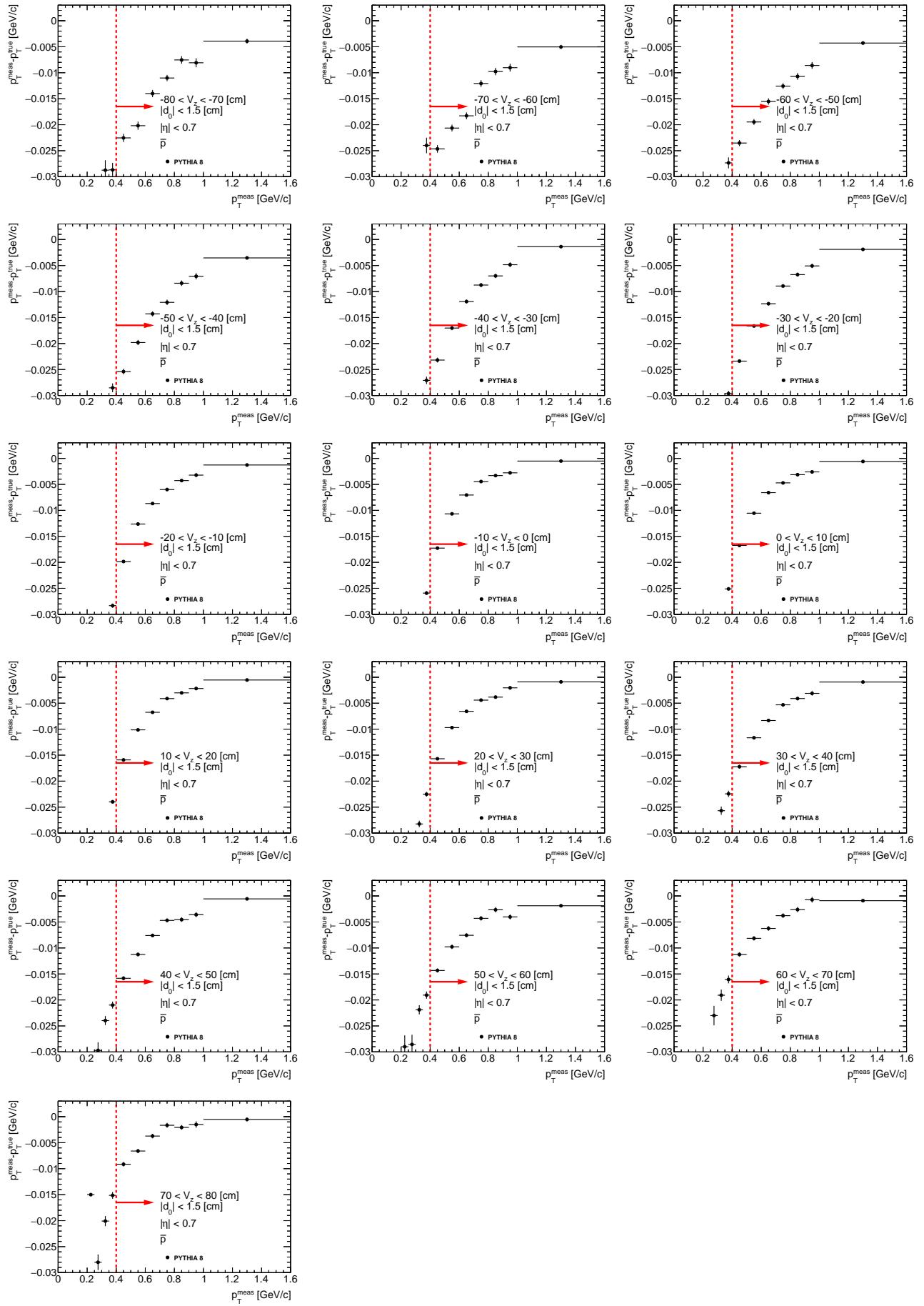


Figure C.6: Energy loss correction $p_T^{meas} - p_T^{true}$ for p as a function of reconstructed transverse momentum p_T^{meas} ($|\eta| < 0.7$) in single z -vertex bin whose range is given on each plot. Red lines and arrows indicate region accepted in analyses.

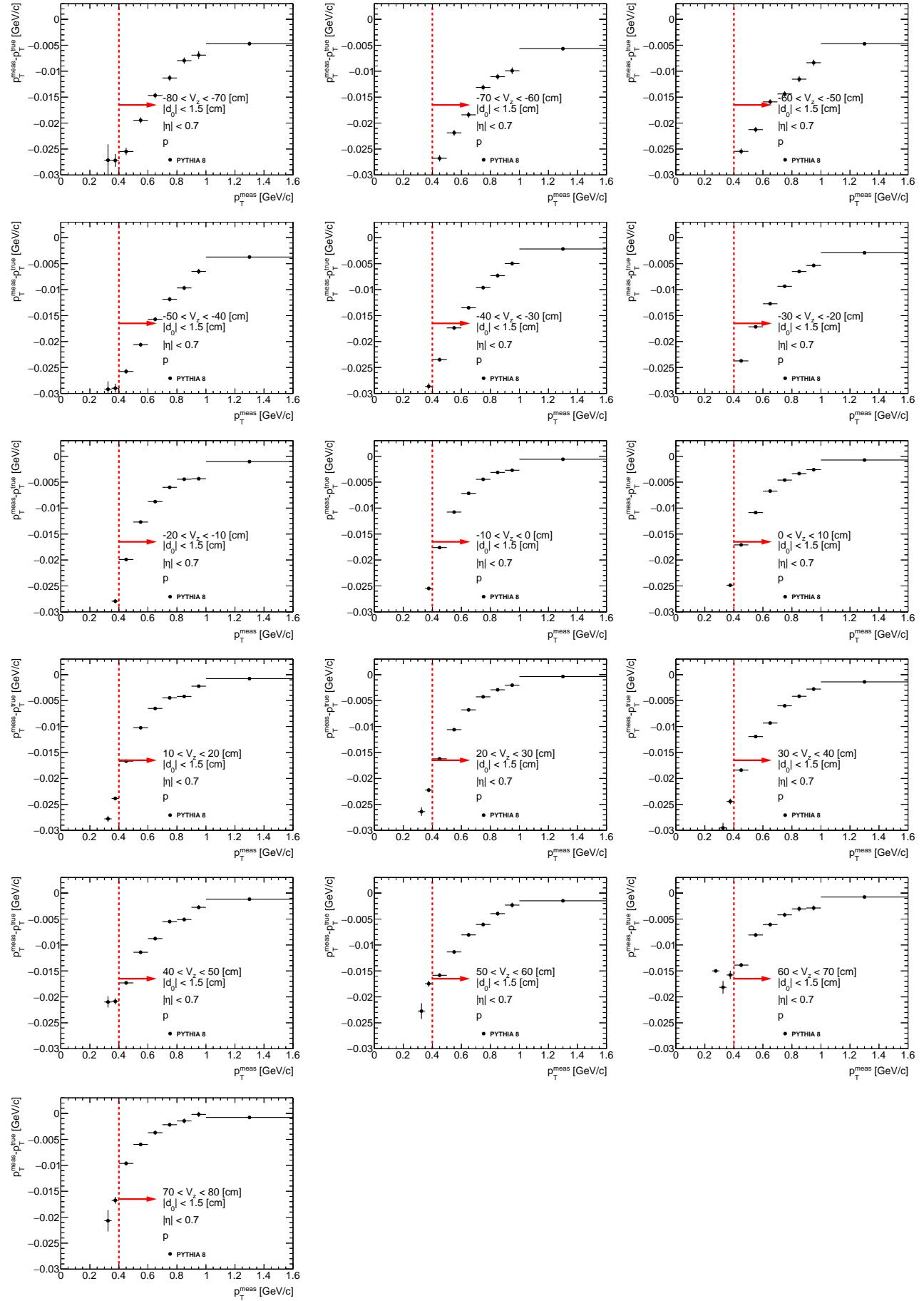


Figure C.7: Energy loss correction $p_T^{meas} - p_T^{true}$ for negative particles as a function of reconstructed transverse momentum p_T^{meas} ($|\eta| < 0.7$) in single z -vertex bin whose range is given on each plot. Red lines and arrows indicate region accepted in analyses.

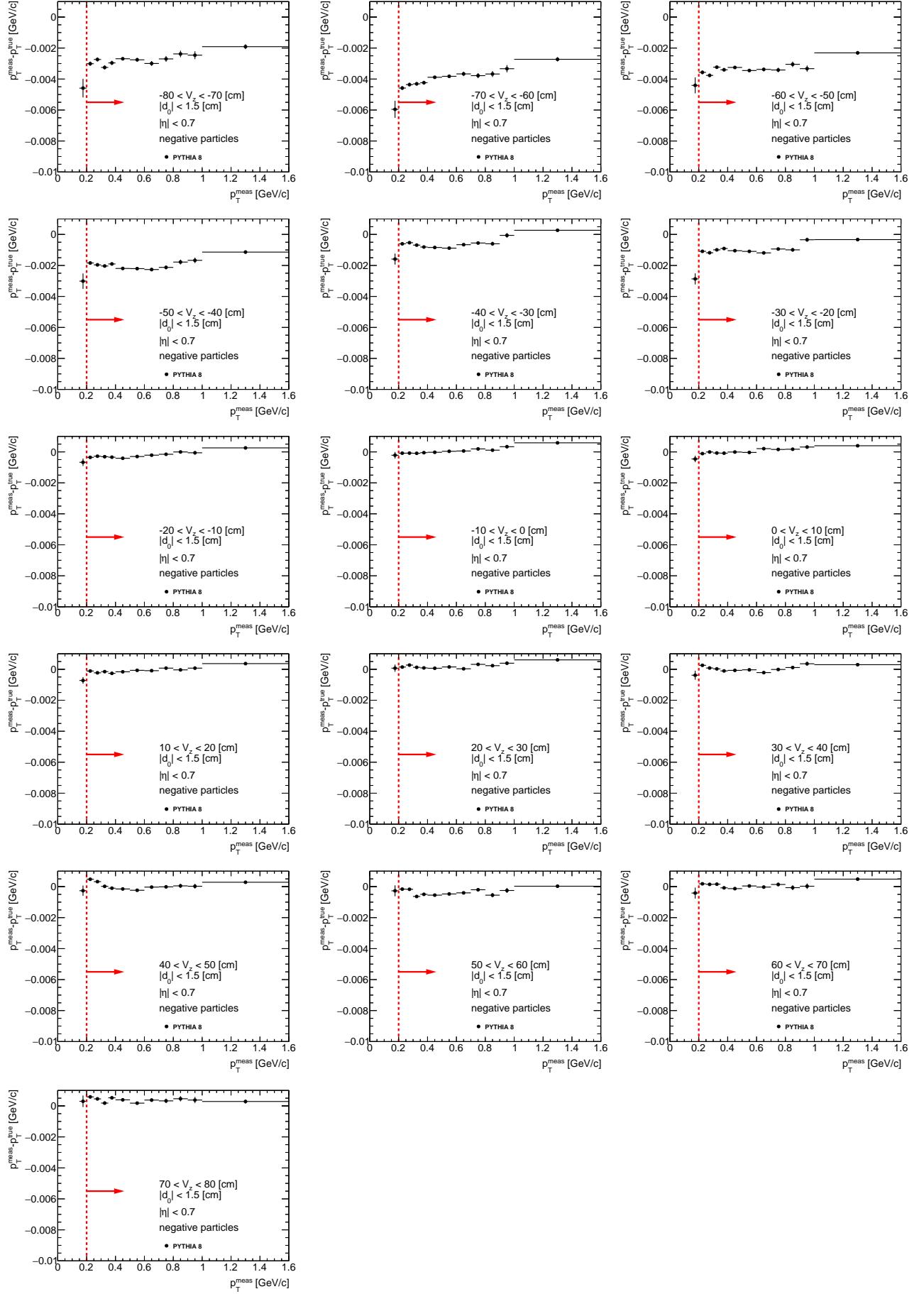


Figure C.8: Energy loss correction $p_T^{meas} - p_T^{true}$ for positive particles as a function of reconstructed transverse momentum p_T^{meas} ($|\eta| < 0.7$) in single z -vertex bin whose range is given on each plot. Red lines and arrows indicate region accepted in analyses.

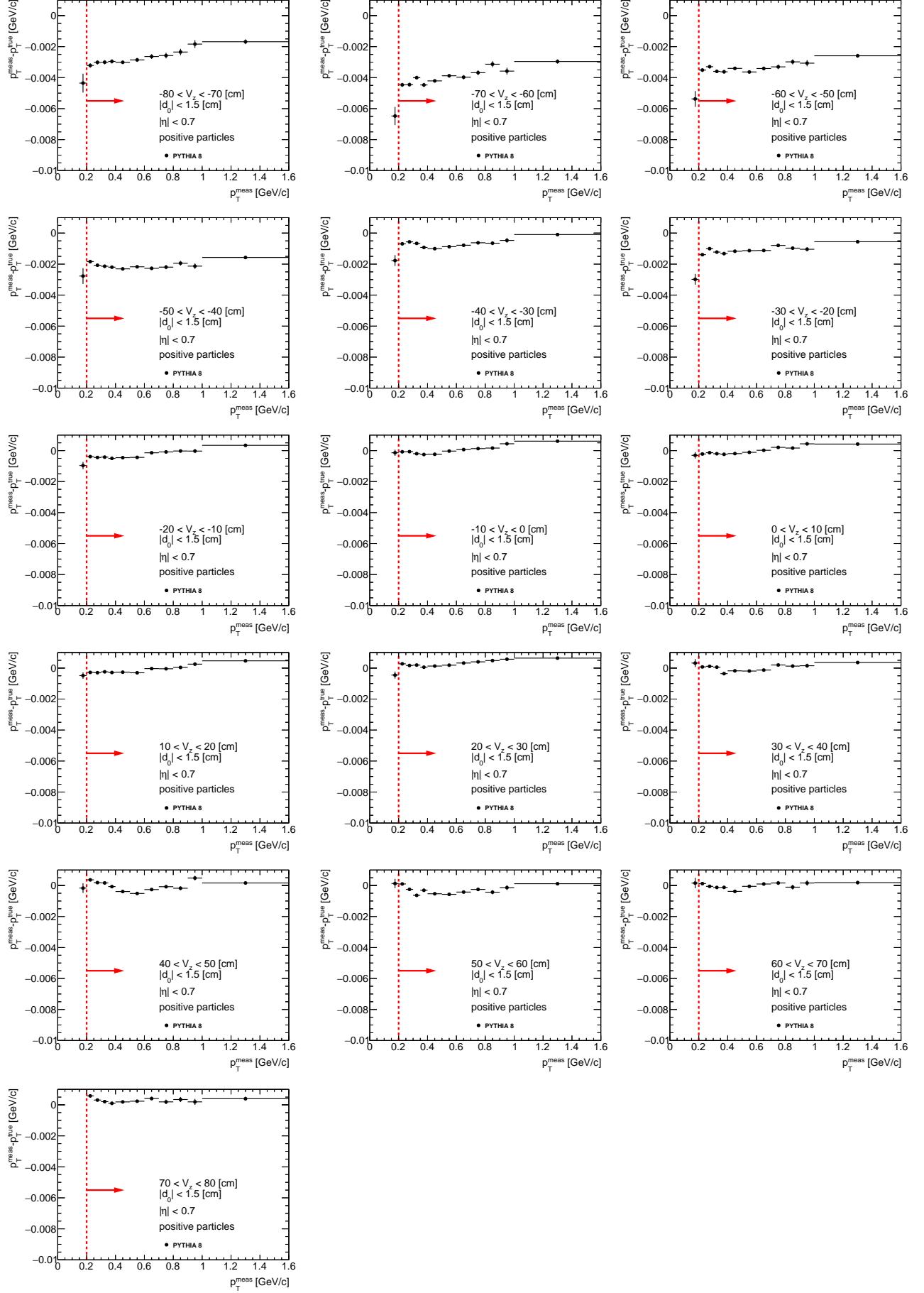
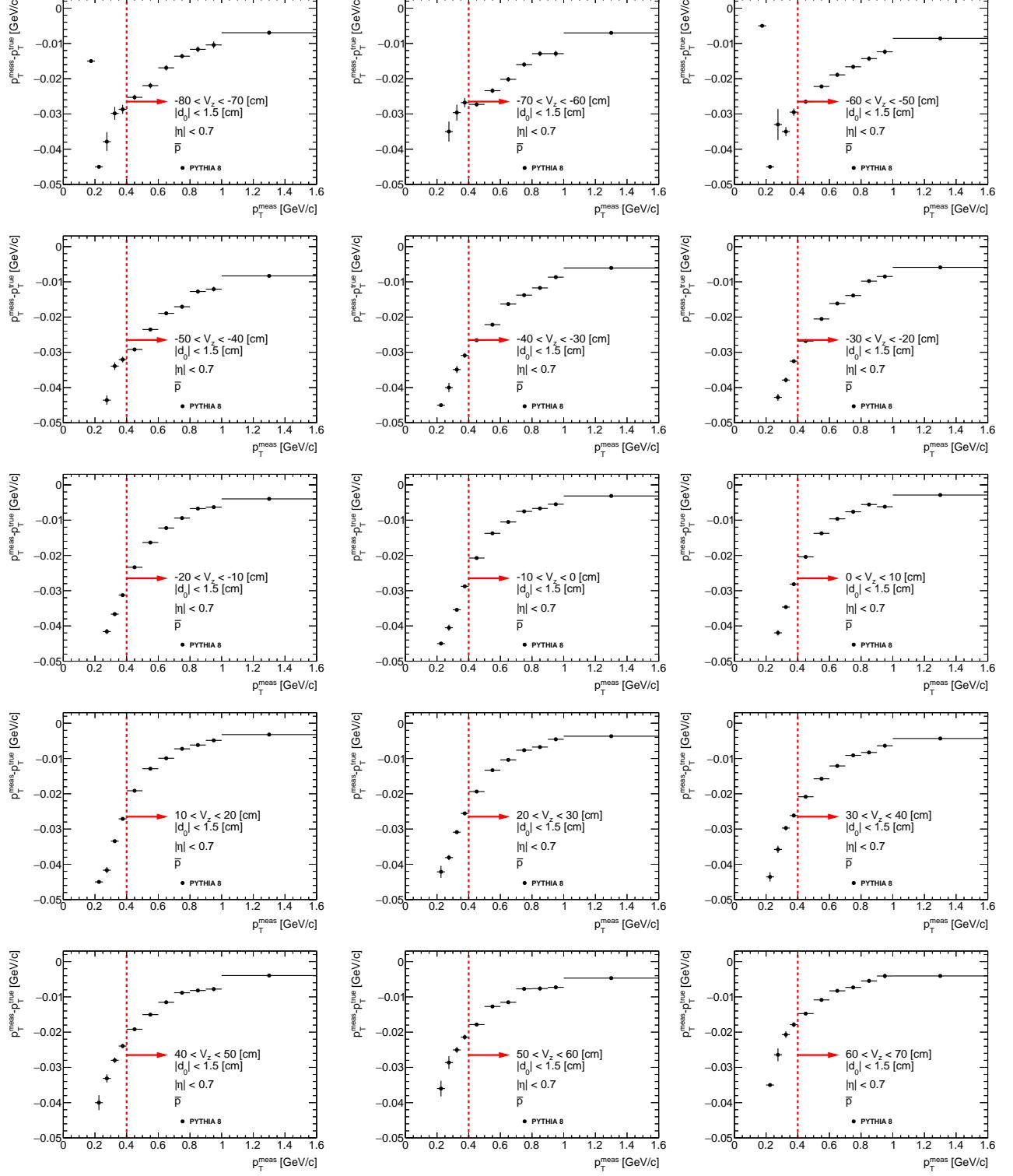


Figure C.9: Energy loss correction $p_T^{meas} - p_T^{true}$ for \bar{p} as a function of reconstructed global track transverse momentum p_T^{meas} ($|\eta| < 0.7$) in single z -vertex bin whose range is given on each plot. Red lines and arrows indicate region accepted in analyses. One may need the energy loss correction for reconstructed global proton and antiproton tracks to estimate the knock-out proton background. During the reconstruction, global tracks are corrected only for energy losses in TPC, whereas primary tracks have the information of dead material inside and in front of TPC. Since that, there is an offset of about 4 MeV for global tracks.



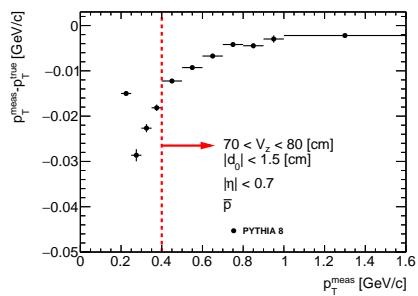
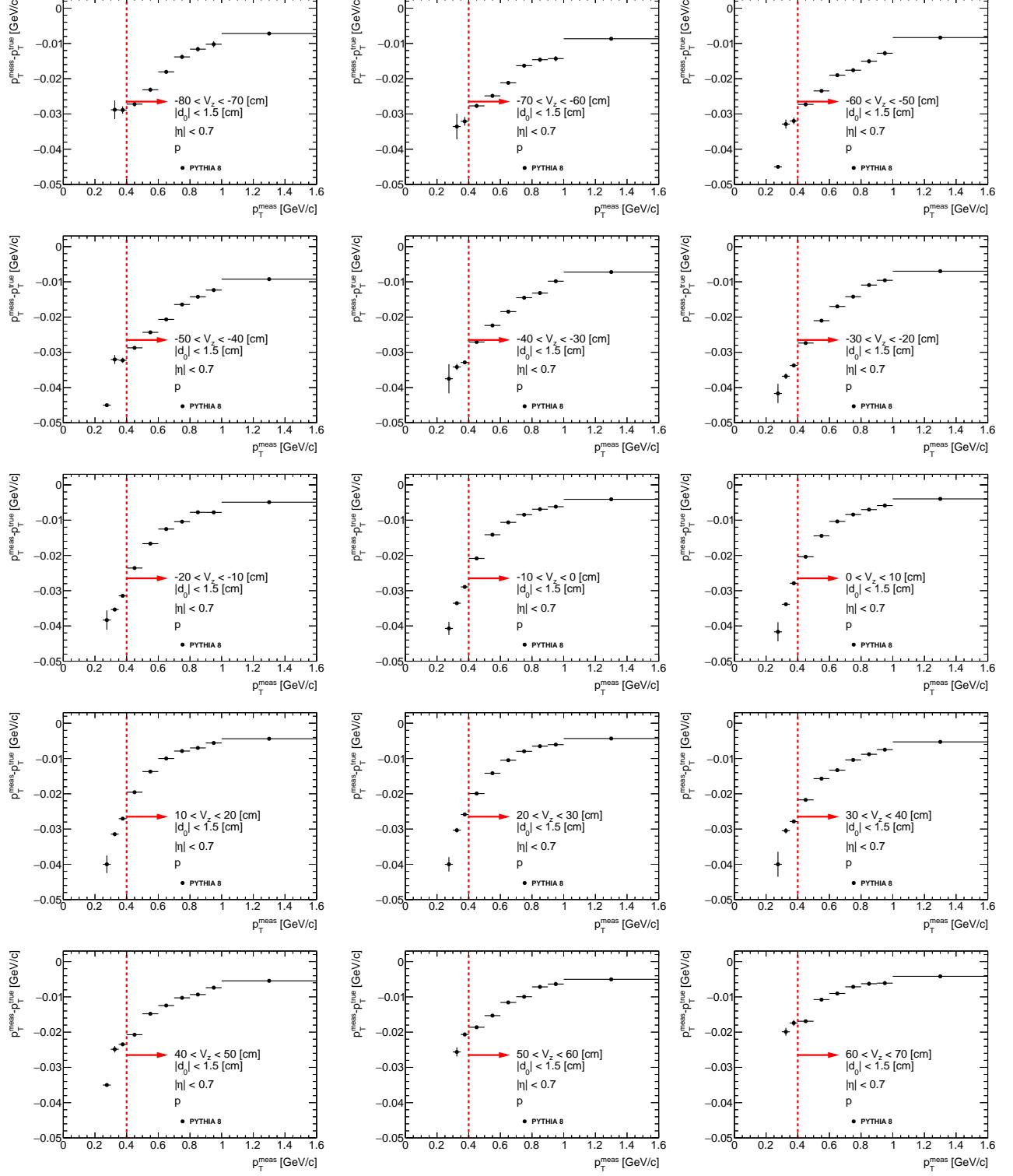
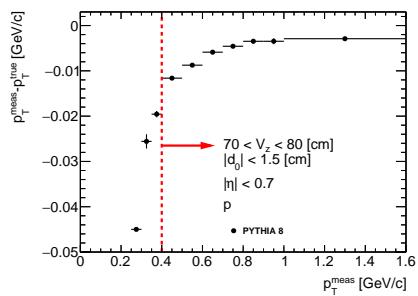


Figure C.10: Energy loss correction $p_T^{meas} - p_T^{true}$ for p as a function of reconstructed global track transverse momentum p_T^{meas} ($|\eta| < 0.7$) in single z -vertex bin whose range is given on each plot. Red lines and arrows indicate region accepted in analyses. One may need the energy loss correction for reconstructed global proton and antiproton tracks to estimate the knock-out proton background. During the reconstruction, global tracks are corrected only for energy losses in TPC, whereas primary tracks have the information of dead material inside and in front of TPC. Since that, there is an offset of about 4 MeV for global tracks.

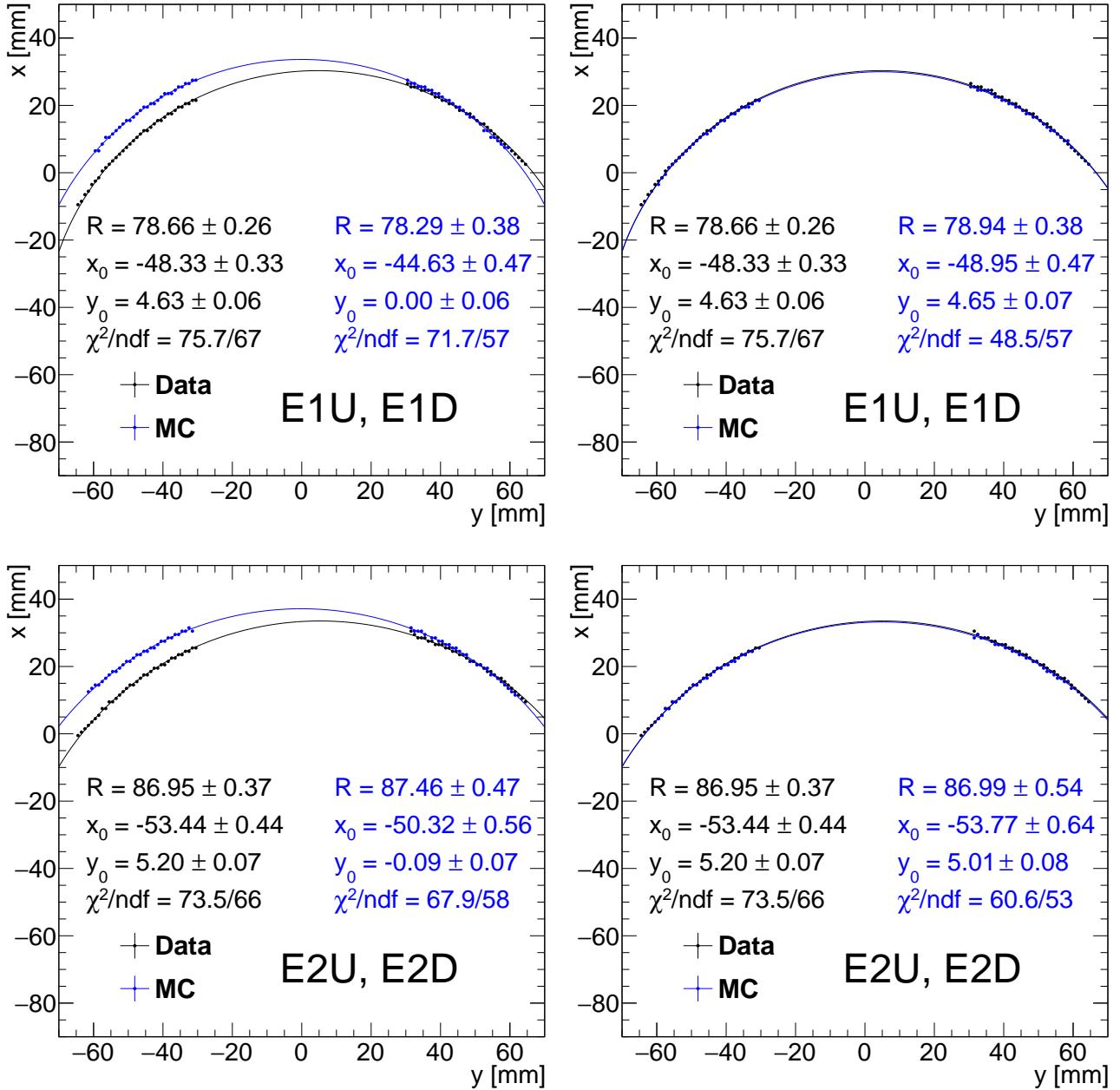


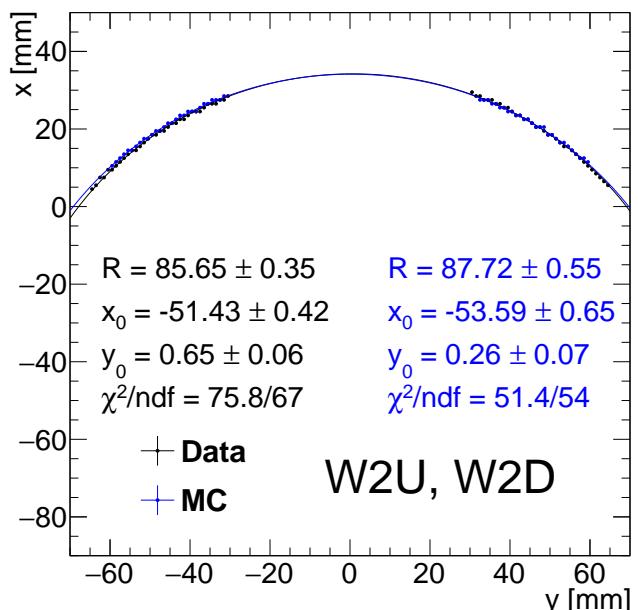
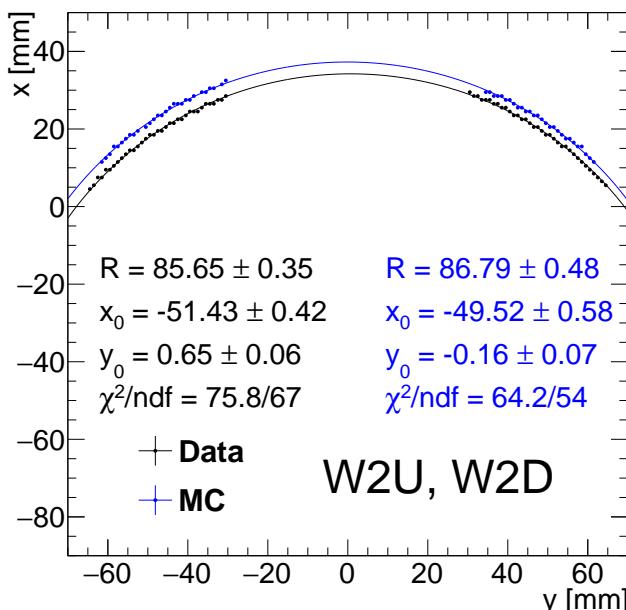
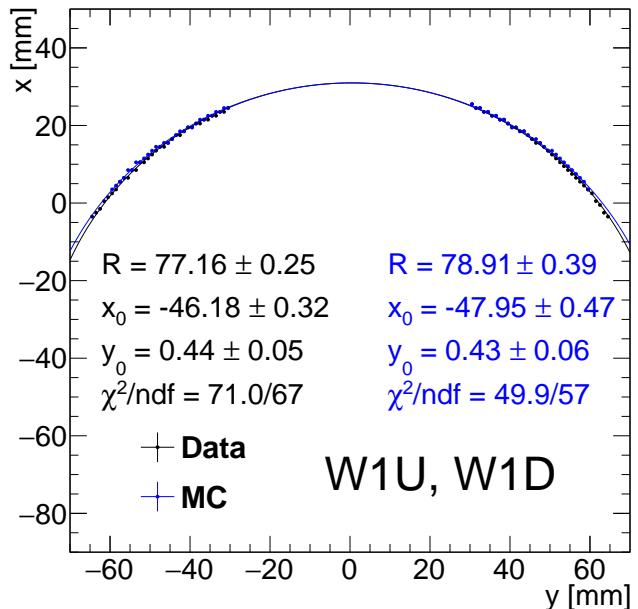
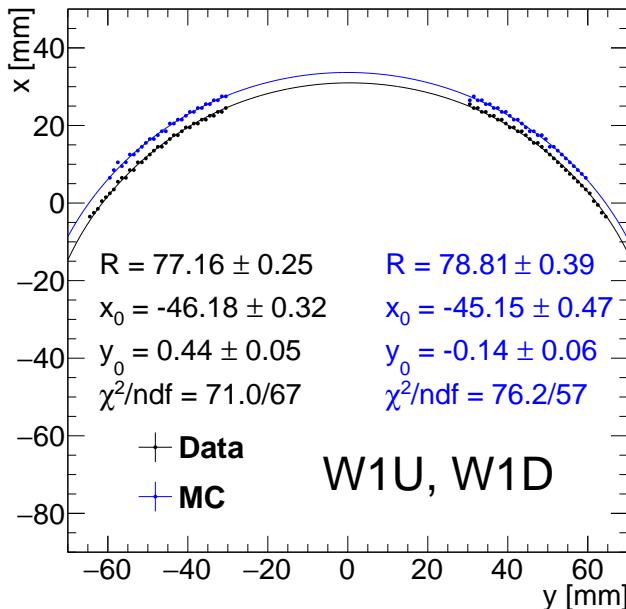


Appendix D

Apertures tuning in Geant4 simulation

Figure D.1: The DX aperture envelopes fitted with circles before (left) and after (right) the DX offsets introduced in the Geant4 geometry.

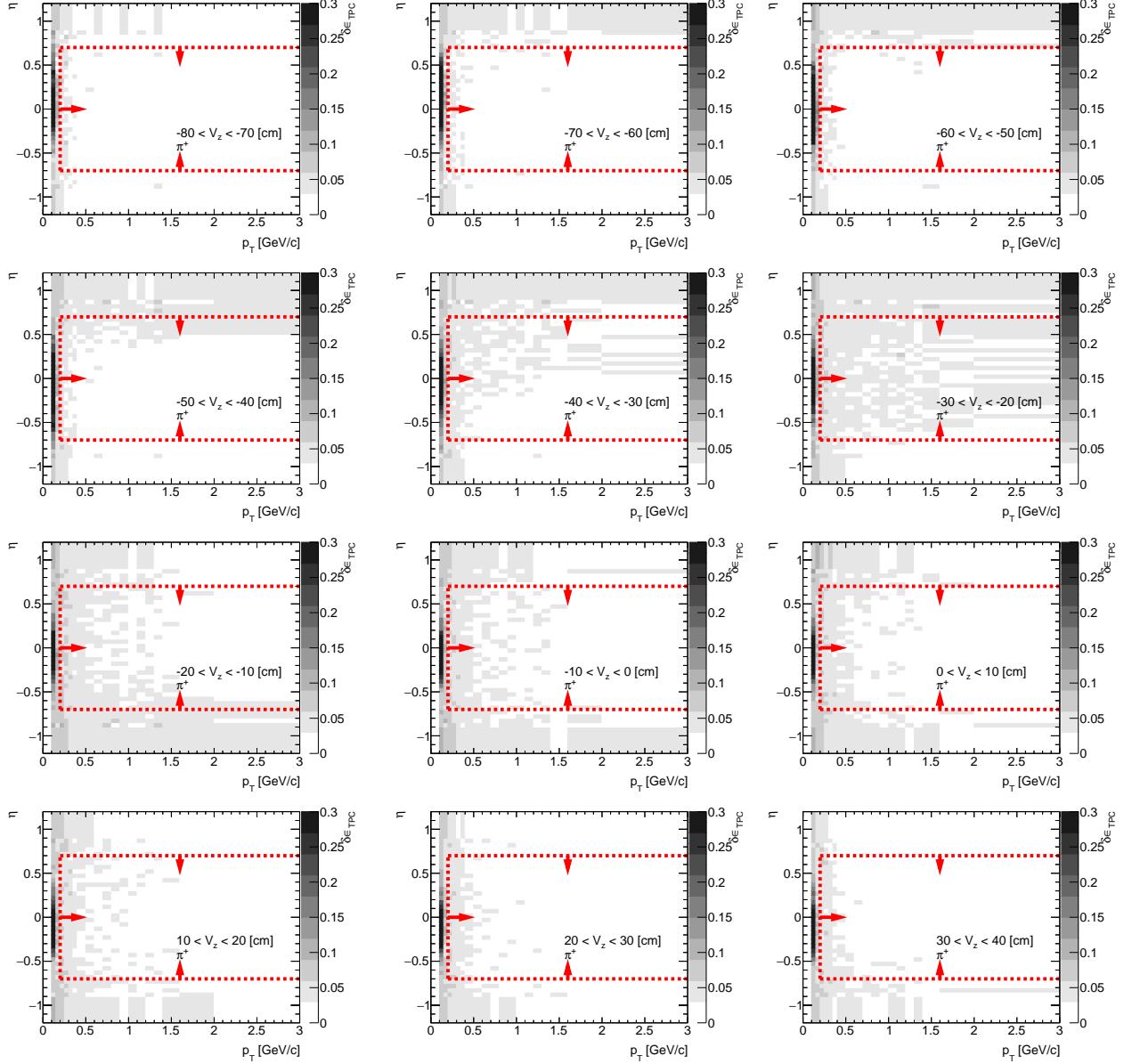




Appendix E

Dead material effect on TPC track reconstruction efficiency

Figure E.1: The amount of lost π^- due to the interaction with dead material in front of TPC in CD MC sample. Each plot represents the fraction of lost π^+ , $\delta\epsilon_{TPC}$ (z -axis), as a function of true particle pseudorapidity η (y -axis) and transverse momentum p_T (x -axis) in single z -vertex bin.



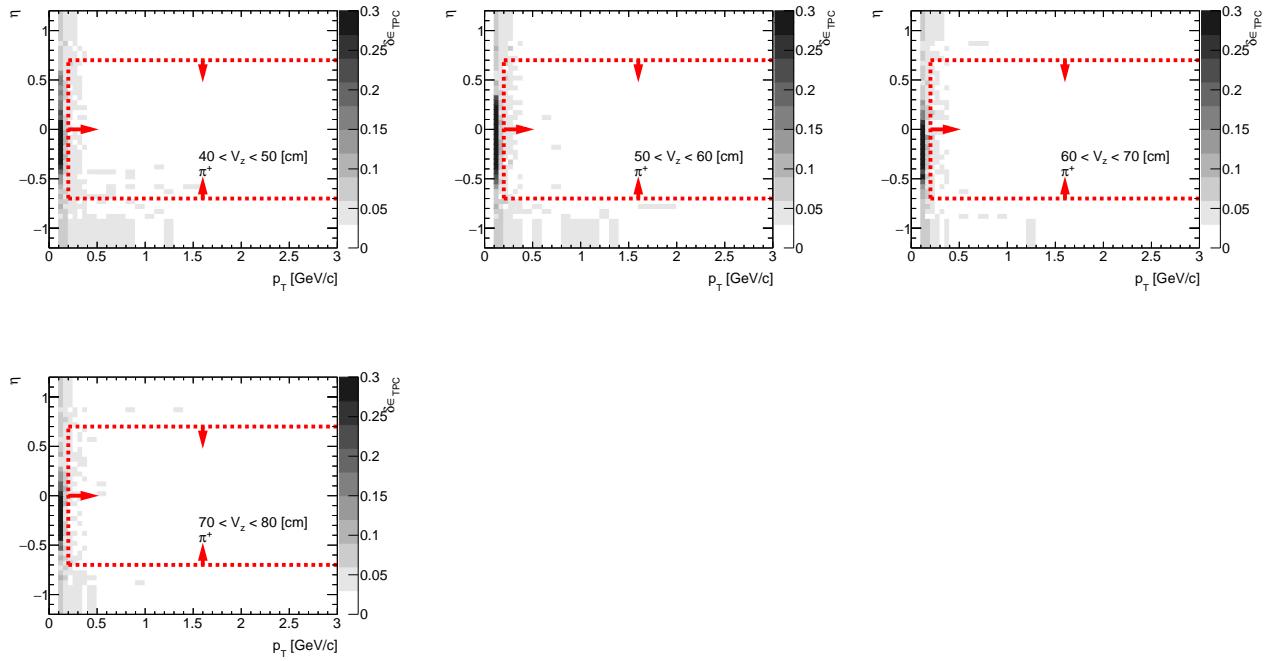


Figure E.2: The amount of lost K^- due to the interaction with dead material in front of TPC in CD MC sample. Each plot represents the fraction of lost K^- , $\delta\epsilon_{TPC}$ (z -axis), as a function of true particle pseudorapidity η (y -axis) and transverse momentum p_T (x -axis) in single z -vertex bin.

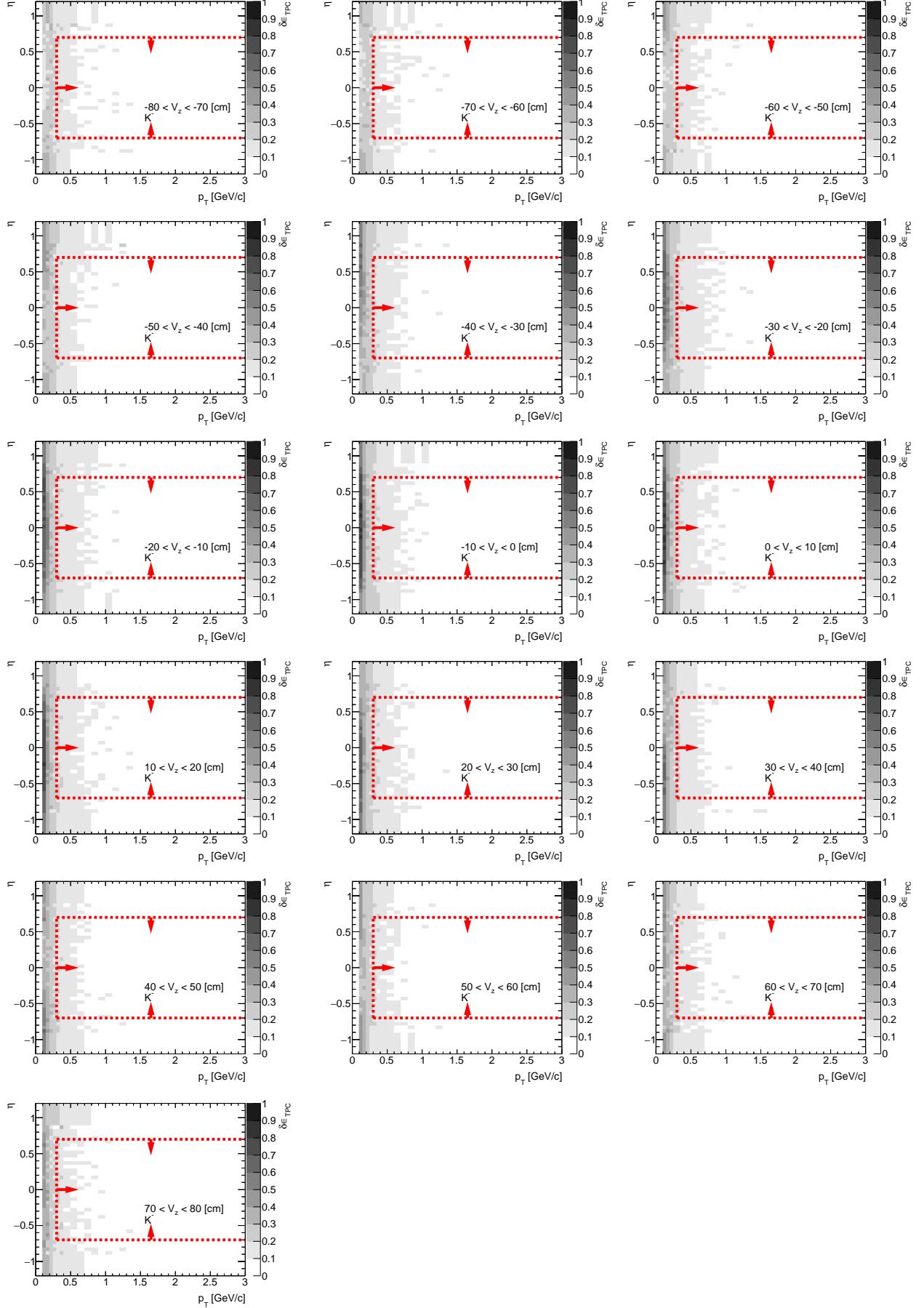


Figure E.3: The amount of lost K^+ due to the interaction with dead material in front of TPC in CD MC sample. Each plot represents the fraction of lost K^+ , $\delta\epsilon_{TPC}$ (z -axis), as a function of true particle pseudorapidity η (y -axis) and transverse momentum p_T (x -axis) in single z -vertex bin.

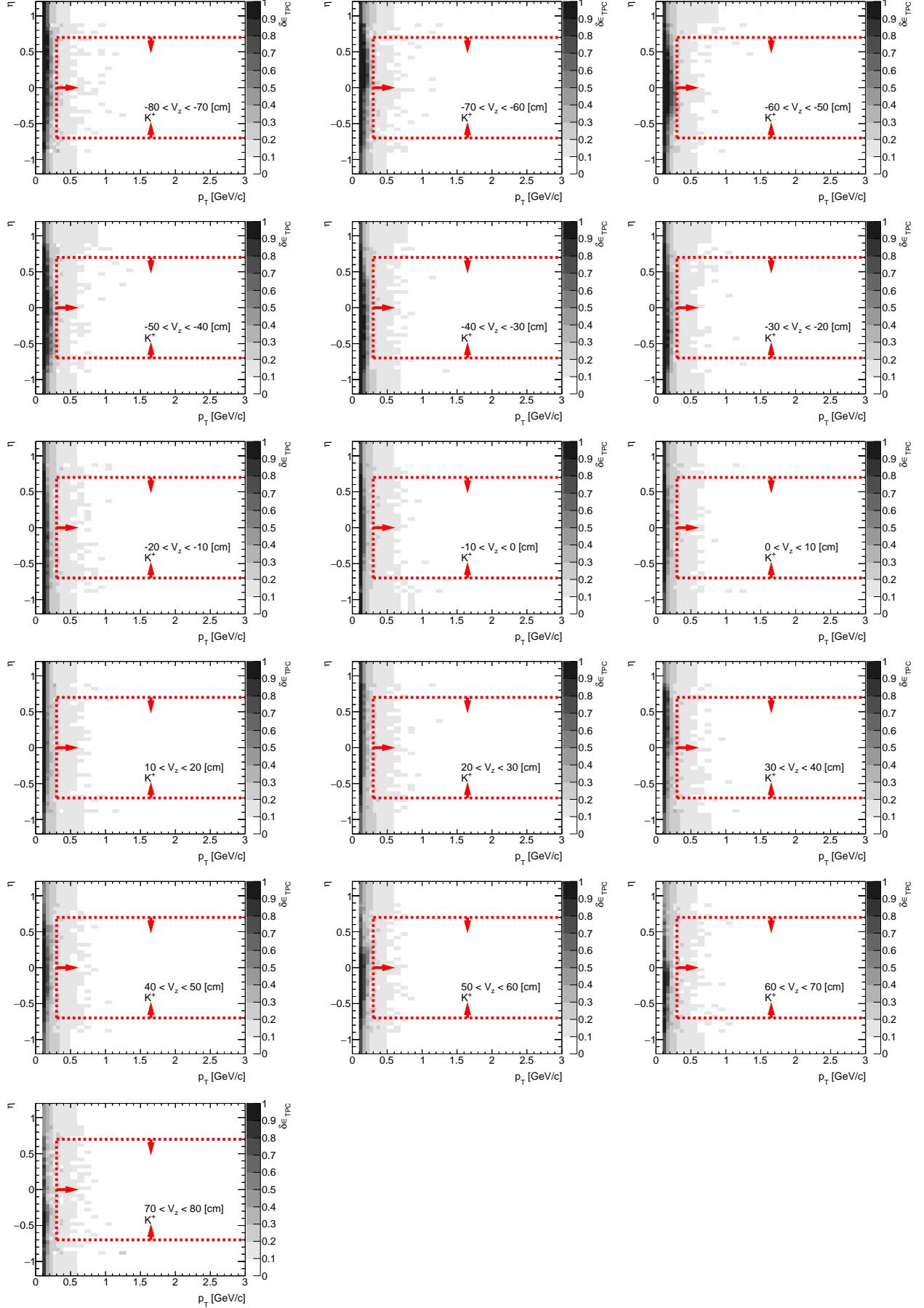


Figure E.4: The amount of lost \bar{p} due to the interaction with dead material in front of TPC in CD MC sample. Each plot represents the fraction of lost \bar{p} , $\delta\epsilon_{TPC}$ (z -axis), as a function of true particle pseudorapidity η (y -axis) and transverse momentum p_T (x -axis) in single z -vertex bin.

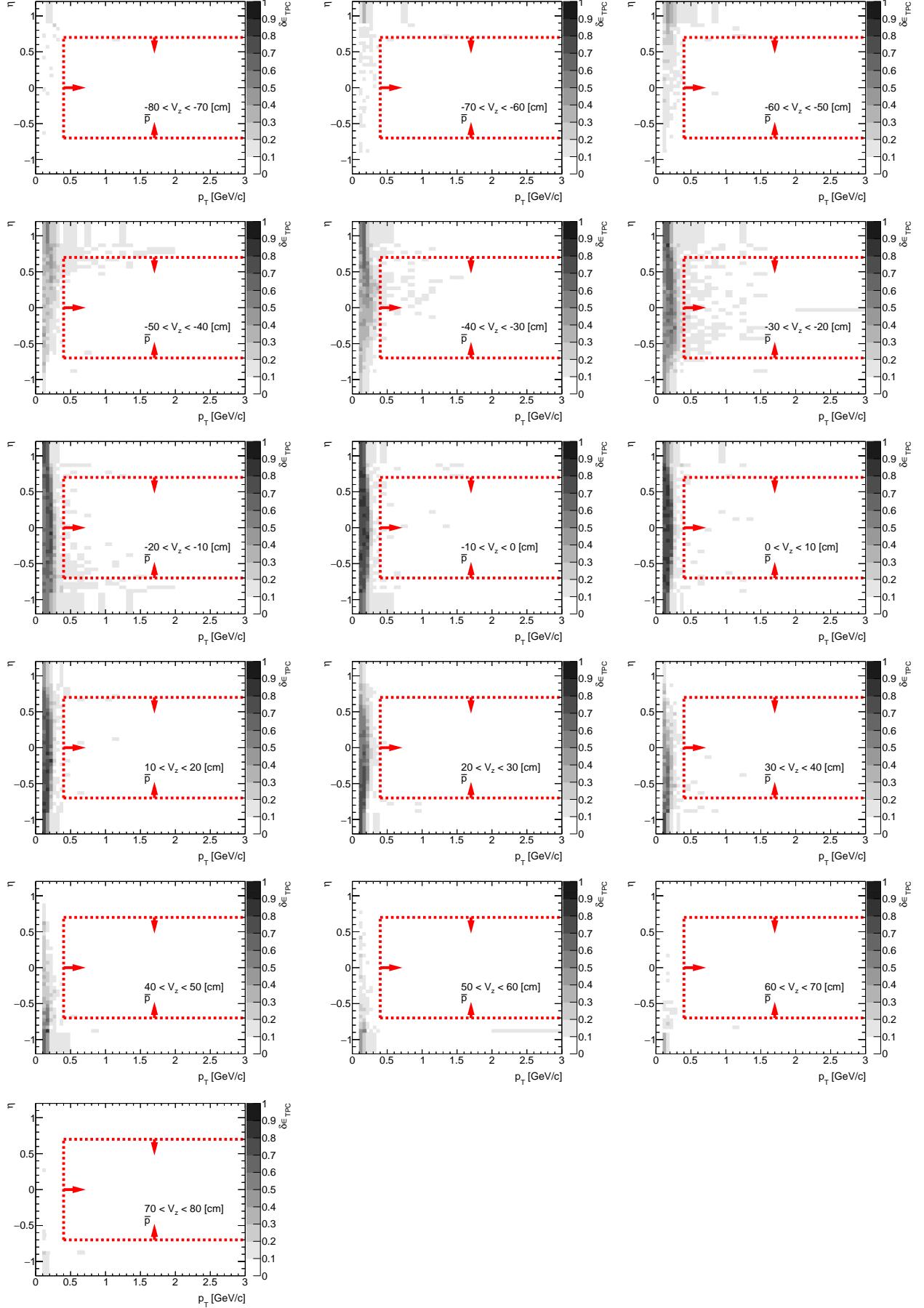


Figure E.5: The amount of lost p due to the interaction with dead material in front of TPC in CD MC sample. Each plot represents the fraction of lost p , $\delta\epsilon_{TPC}$ (z -axis), as a function of true particle pseudorapidity η (y -axis) and transverse momentum p_T (x -axis) in single z -vertex bin.

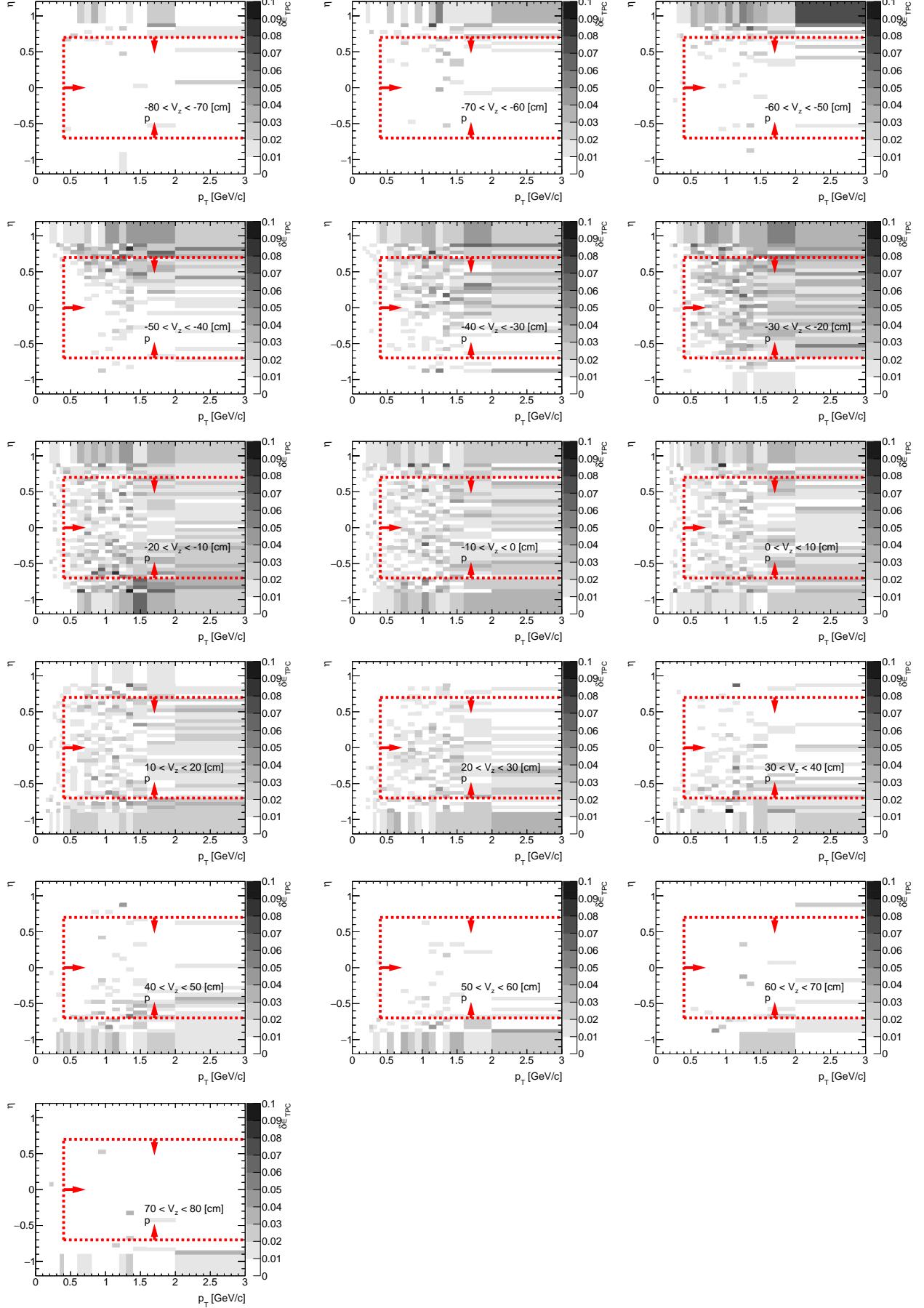


Figure E.6: The amount of lost negative particles due to the interaction with dead material in front of TPC in CD MC sample. Each plot represents the fraction of lost negative particles, $\delta\epsilon_{TPC}$ (z-axis), as a function of true particle pseudorapidity η (y-axis) and transverse momentum p_T (x-axis) in single z -vertex bin.

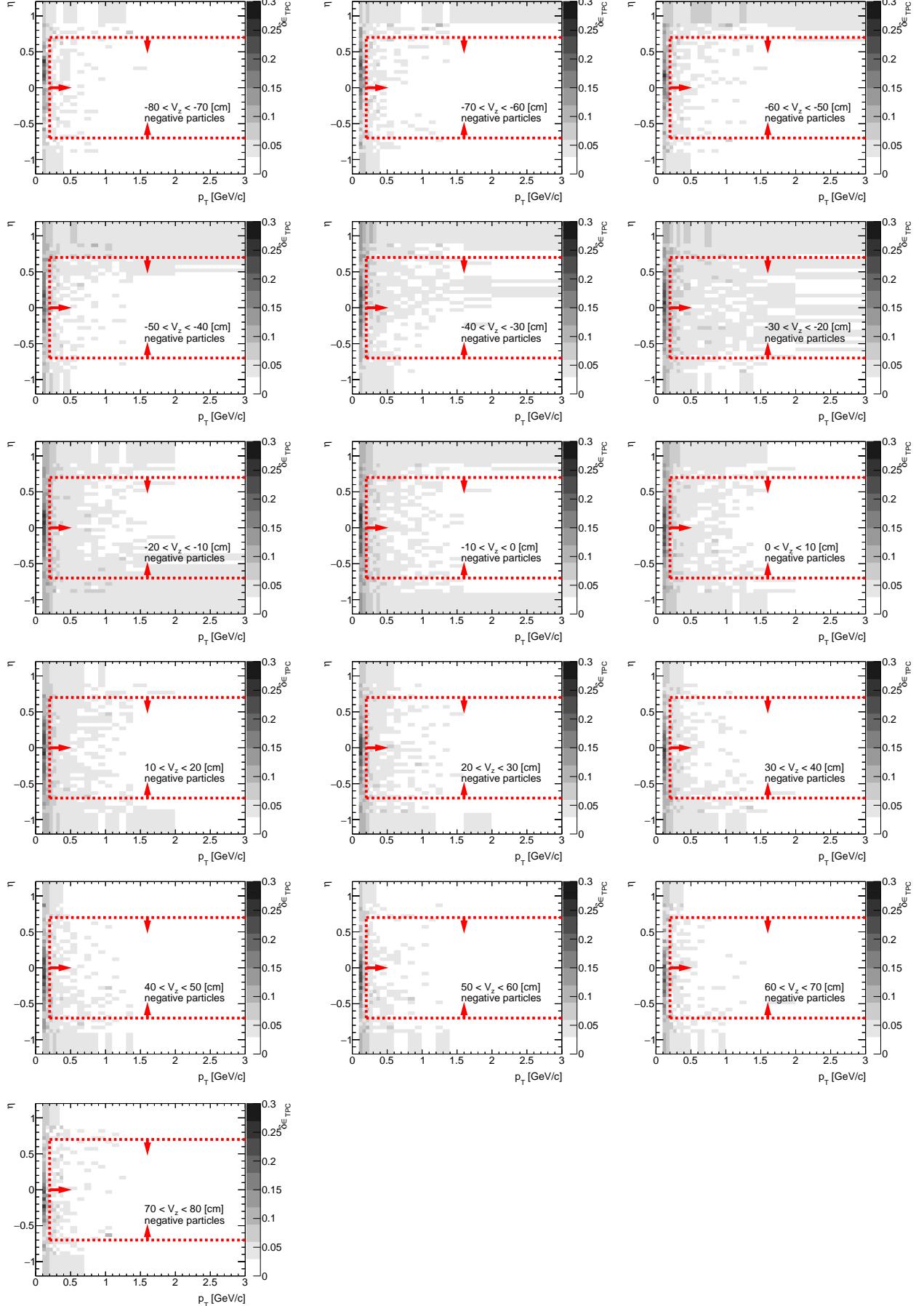


Figure E.7: The amount of lost positive particles due to the interaction with dead material in front of TPC in CD MC sample. Each plot represents the fraction of lost positive particles, $\delta\epsilon_{TPC}$ (z-axis), as a function of true particle pseudorapidity η (y-axis) and transverse momentum p_T (x-axis) in single z -vertex bin.

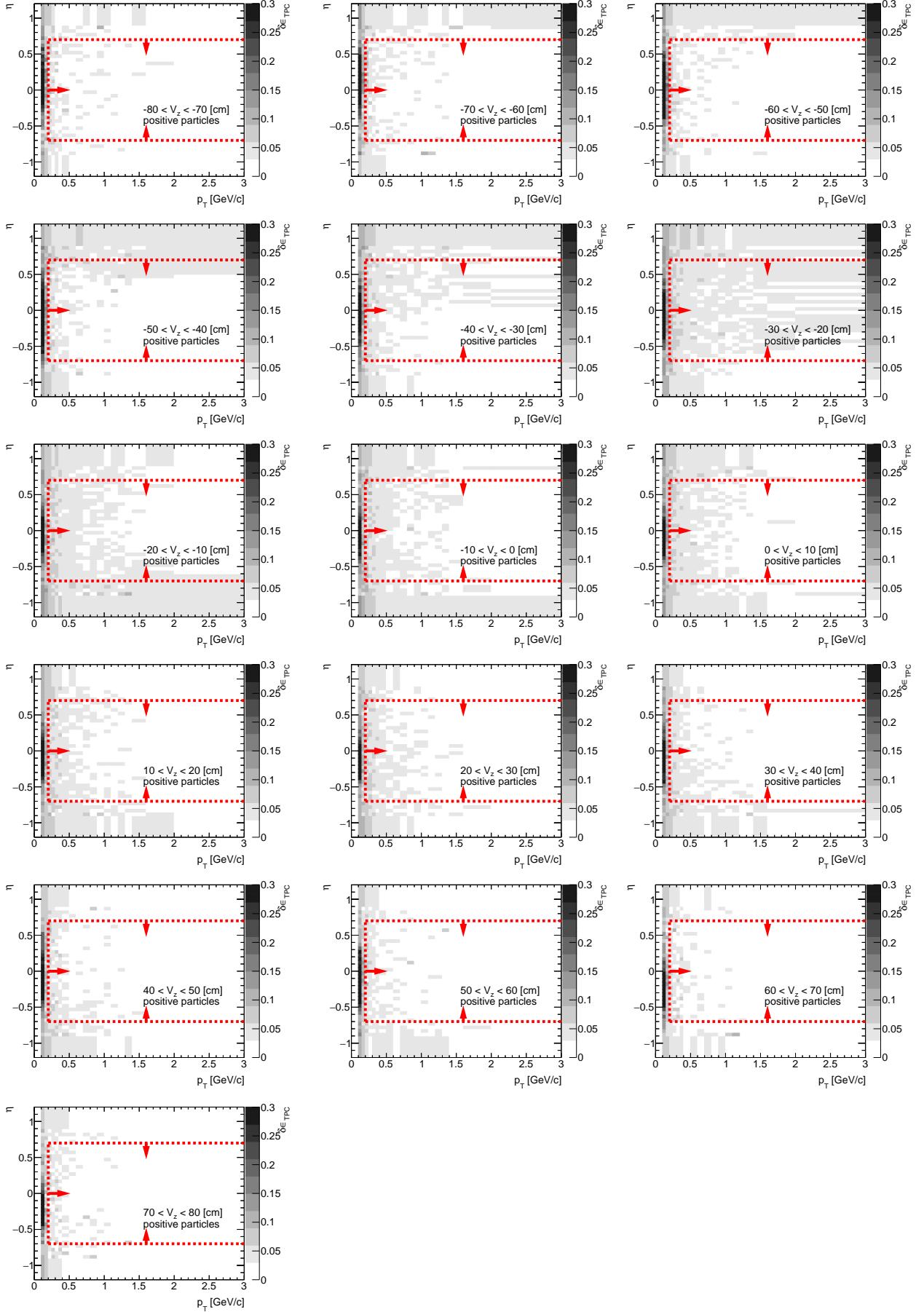


Figure E.8: The amount of lost π^- due to the interaction with dead material in front of TPC in SD MC sample. Each plot represents the fraction of lost π^- , $\delta\epsilon_{TPC}$ (z -axis), as a function of true particle pseudorapidity η (y -axis) and transverse momentum p_T (x -axis) in single z -vertex bin.

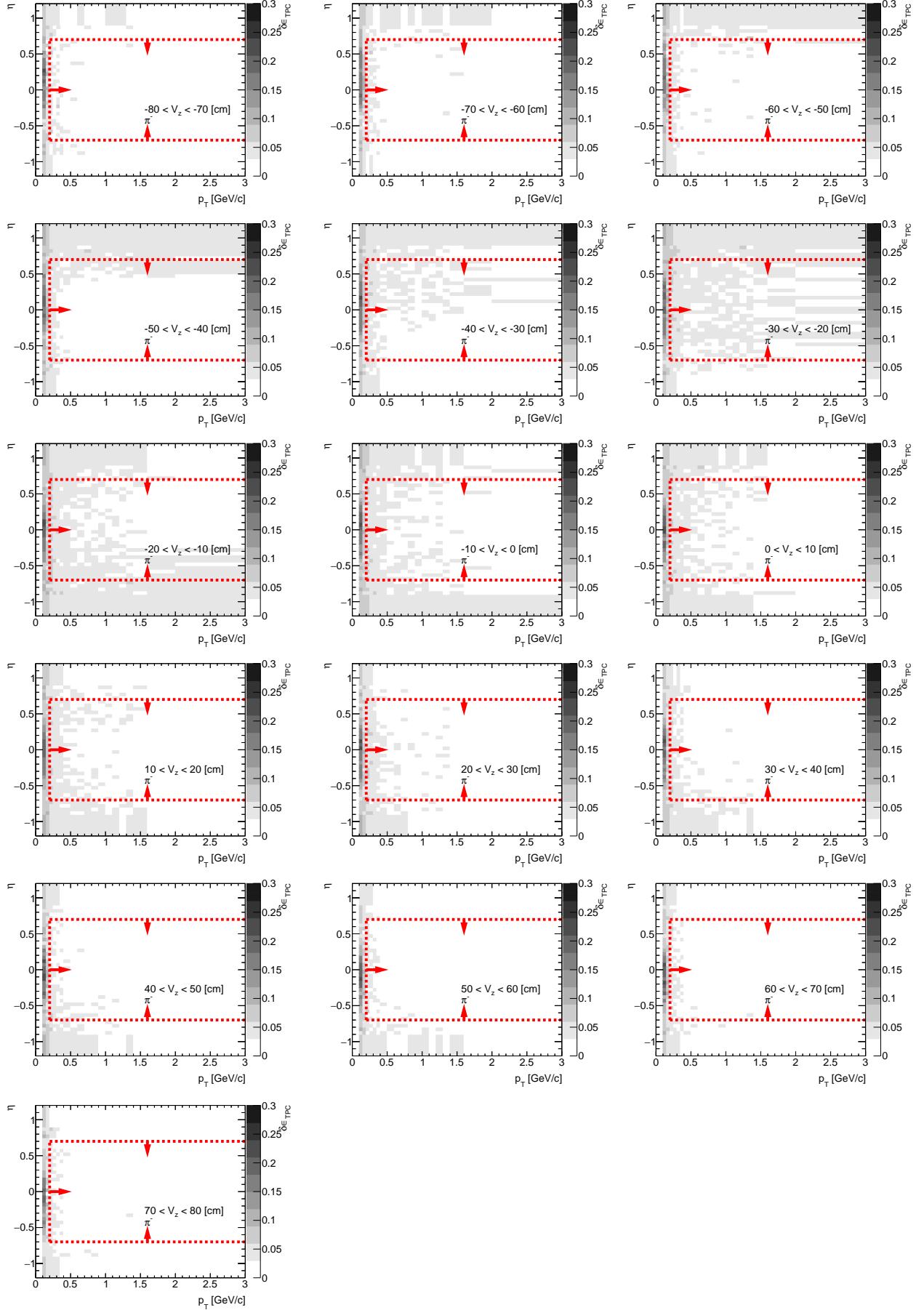


Figure E.9: The amount of lost π^+ due to the interaction with dead material in front of TPC in SD MC sample. Each plot represents the fraction of lost π^+ , $\delta\epsilon_{TPC}$ (z -axis), as a function of true particle pseudorapidity η (y -axis) and transverse momentum p_T (x -axis) in single z -vertex bin.

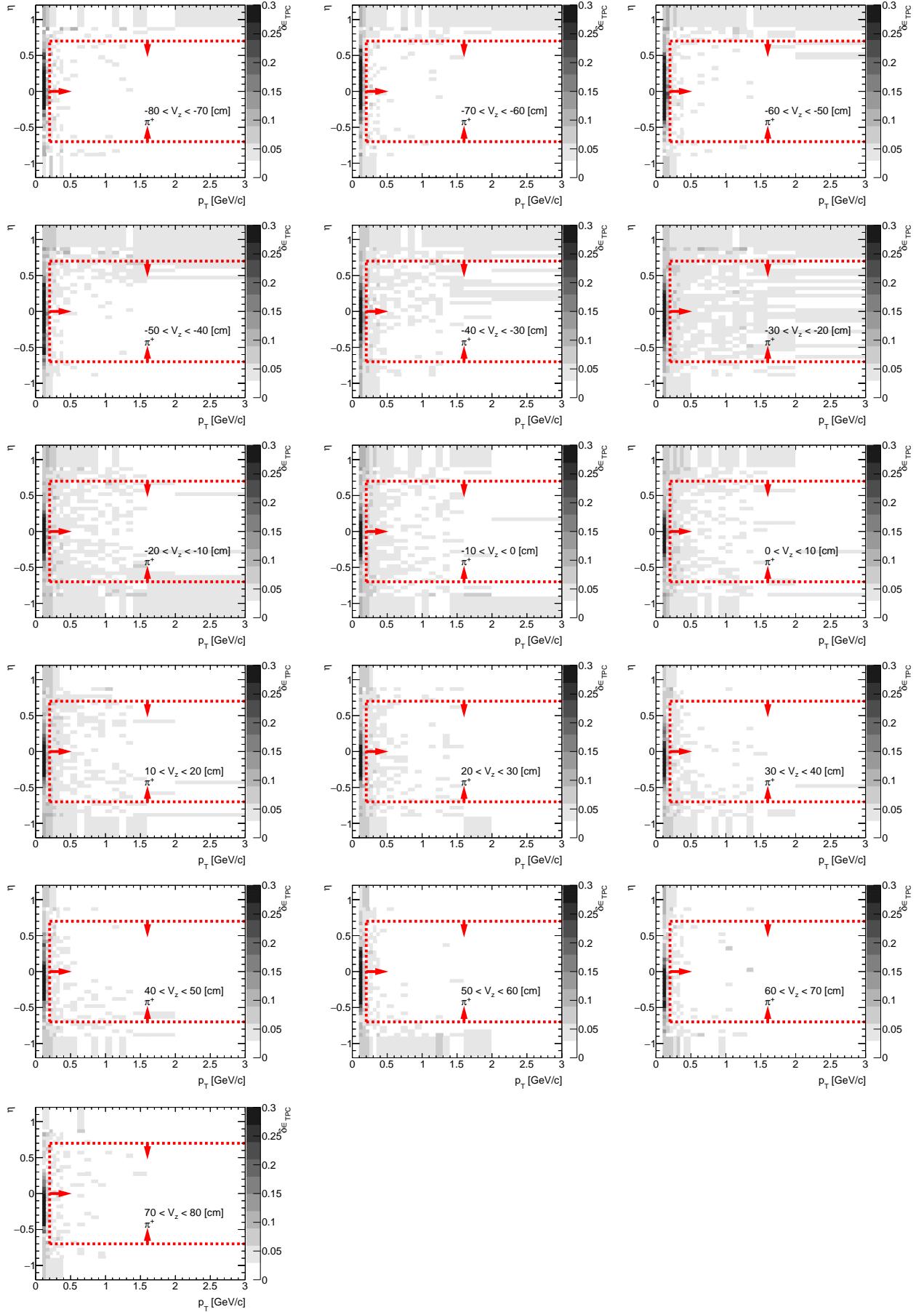


Figure E.10: The amount of lost K^- due to the interaction with dead material in front of TPC in SD MC sample. Each plot represents the fraction of lost K^- , $\delta\epsilon_{TPC}$ (z -axis), as a function of true particle pseudorapidity η (y -axis) and transverse momentum p_T (x -axis) in single z -vertex bin.

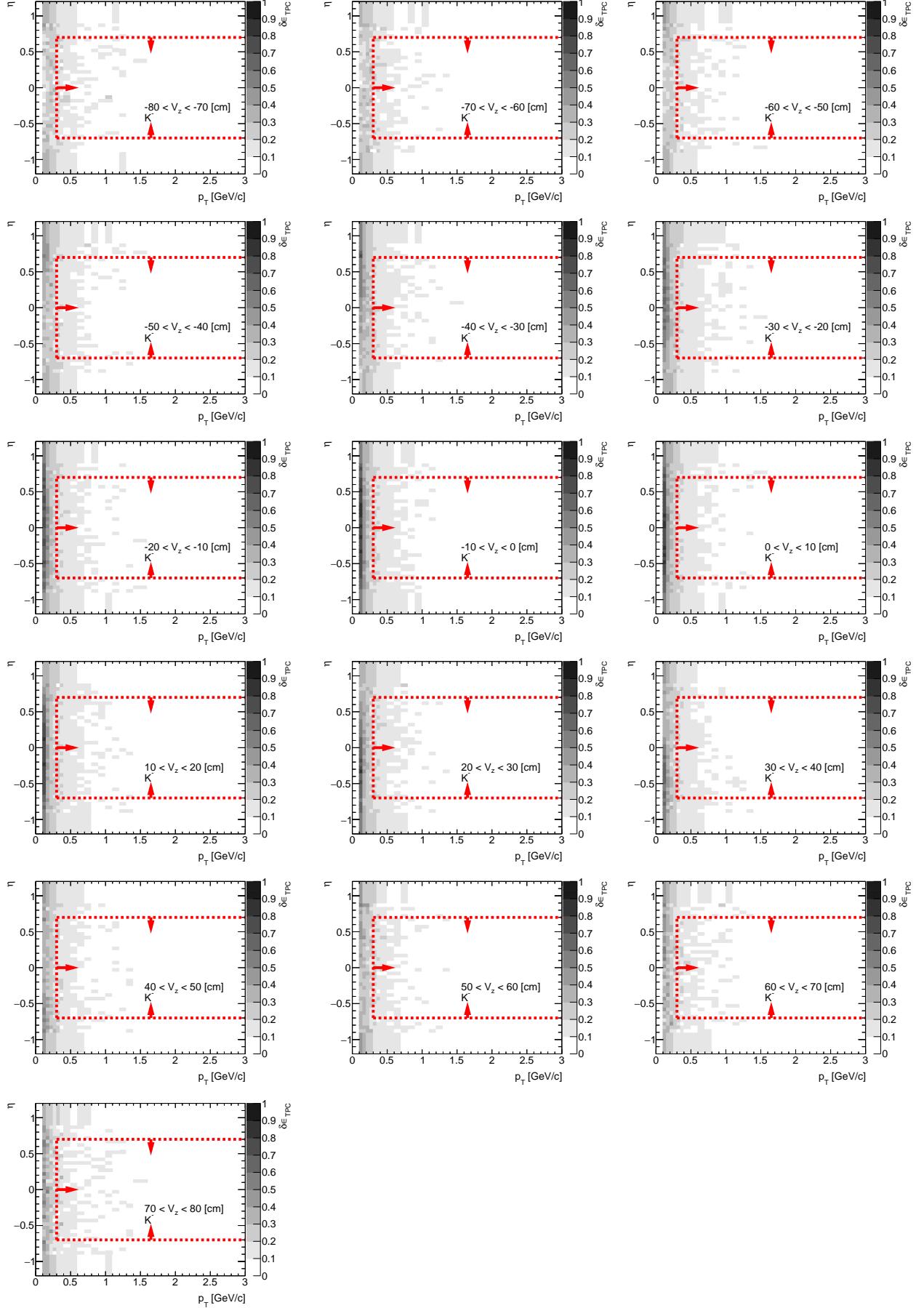


Figure E.11: The amount of lost K^+ due to the interaction with dead material in front of TPC in SD MC sample. Each plot represents the fraction of lost K^+ , $\delta\epsilon_{TPC}$ (z -axis), as a function of true particle pseudorapidity η (y -axis) and transverse momentum p_T (x -axis) in single z -vertex bin.

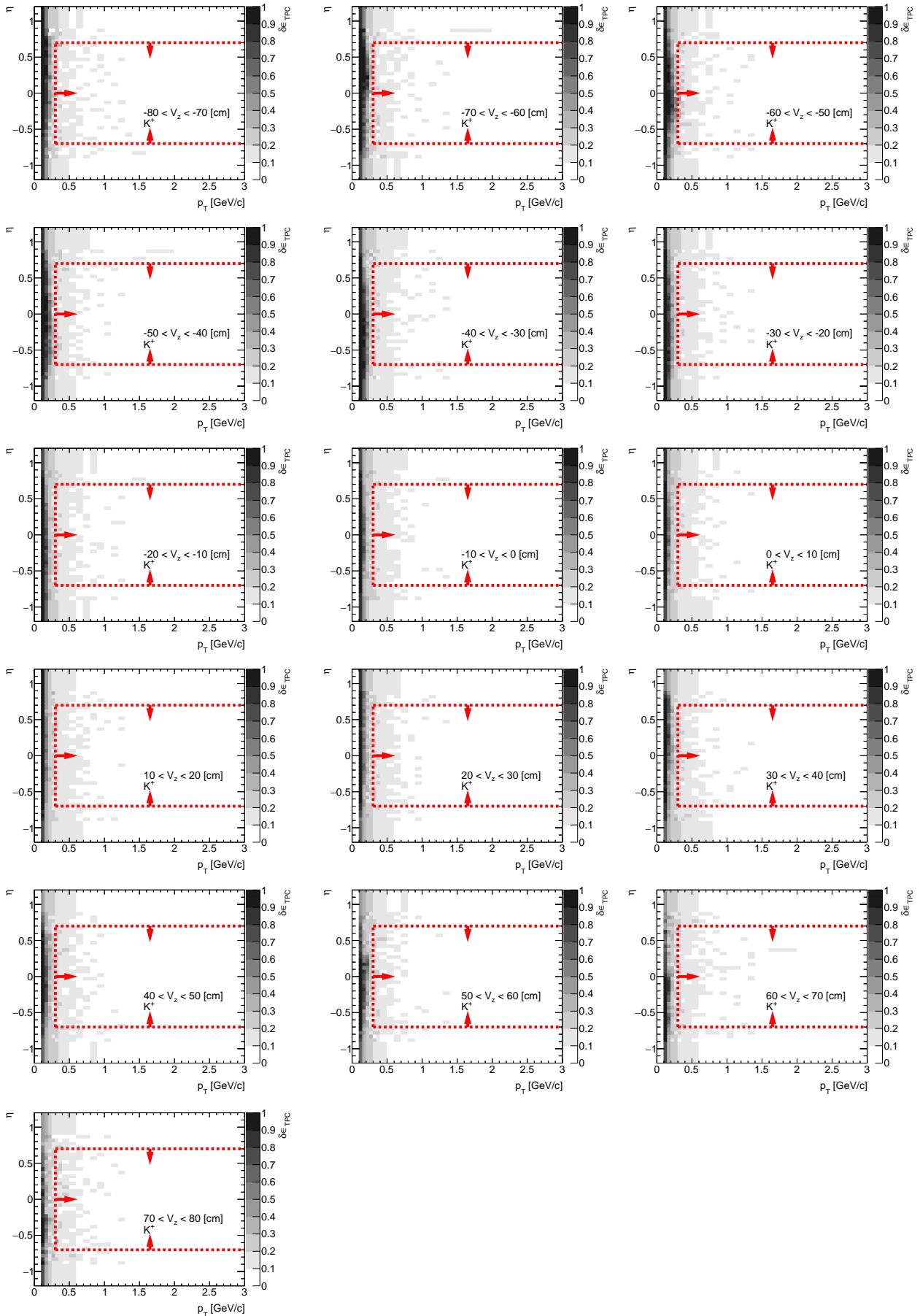


Figure E.12: The amount of lost \bar{p} due to the interaction with dead material in front of TPC in SD MC sample. Each plot represents the fraction of lost \bar{p} , $\delta\epsilon_{TPC}$ (z -axis), as a function of true particle pseudorapidity η (y -axis) and transverse momentum p_T (x -axis) in single z -vertex bin.

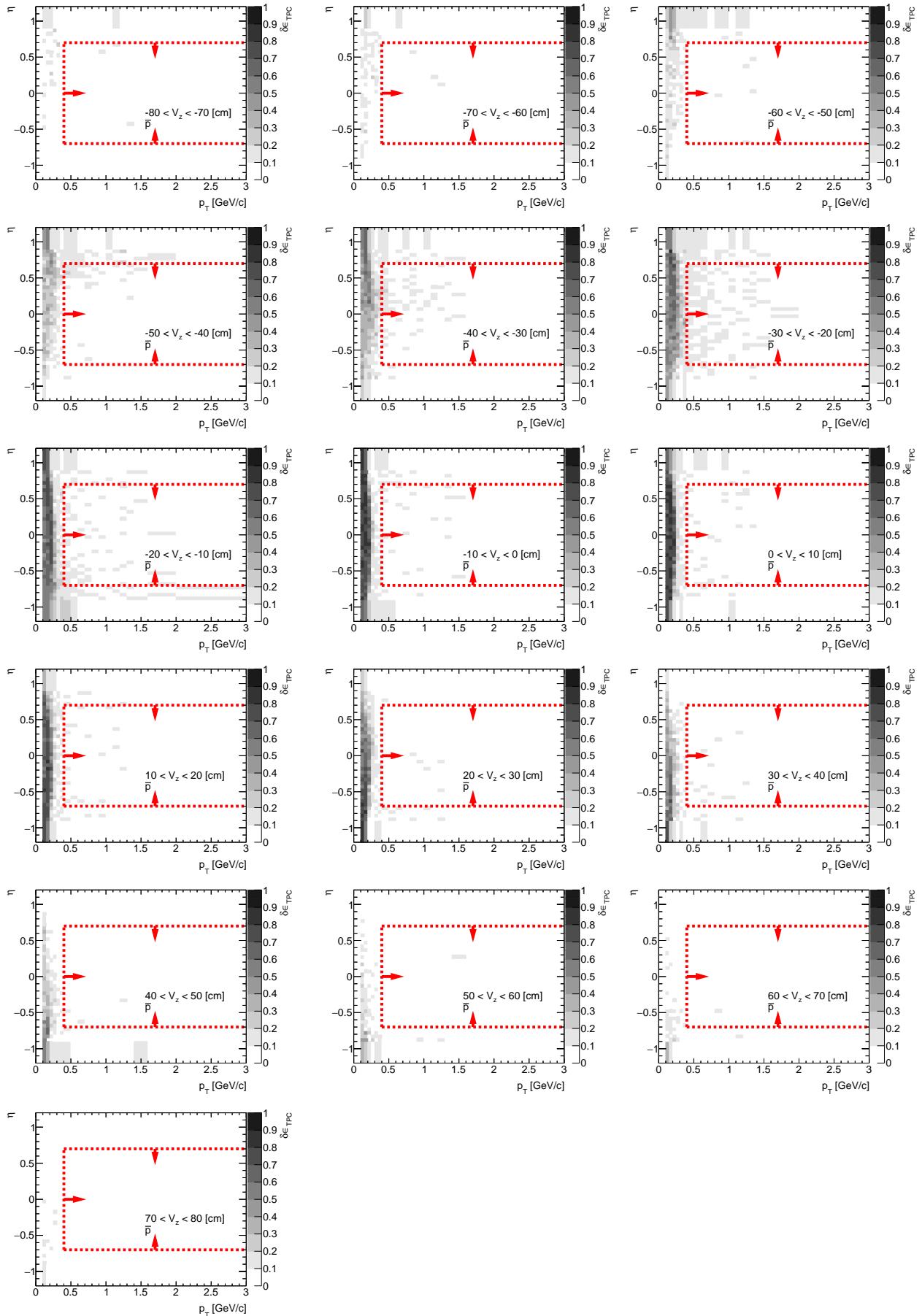


Figure E.13: The amount of lost p due to the interaction with dead material in front of TPC in SD MC sample. Each plot represents the fraction of lost p , $\delta\epsilon_{TPC}$ (z -axis), as a function of true particle pseudorapidity η (y -axis) and transverse momentum p_T (x -axis) in single z -vertex bin.

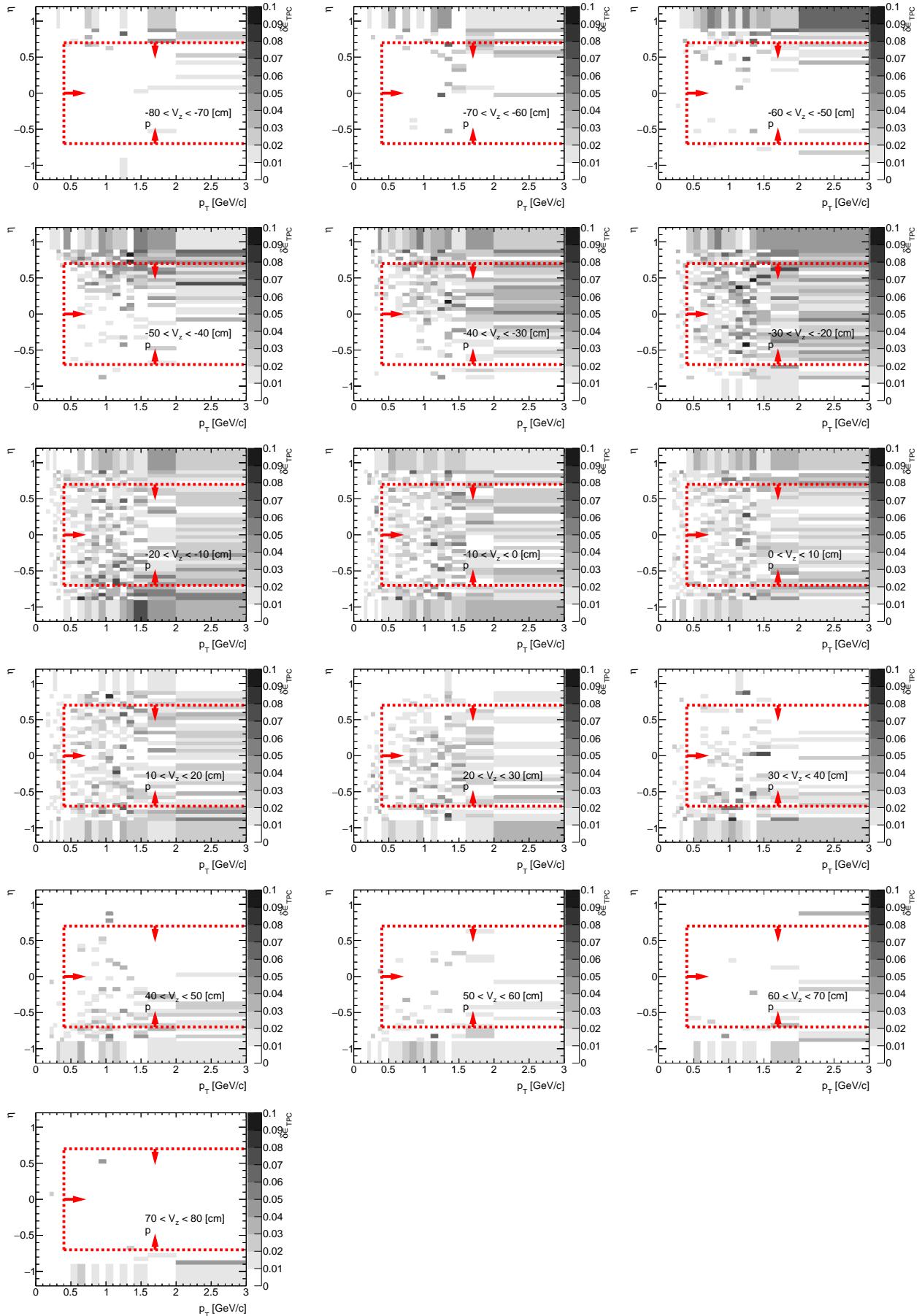


Figure E.14: The amount of lost negative particles due to the interaction with dead material in front of TPC in SD MC sample. Each plot represents the fraction of lost negative particles, $\delta\epsilon_{TPC}$ (z-axis), as a function of true particle pseudorapidity η (y-axis) and transverse momentum p_T (x-axis) in single z -vertex bin.

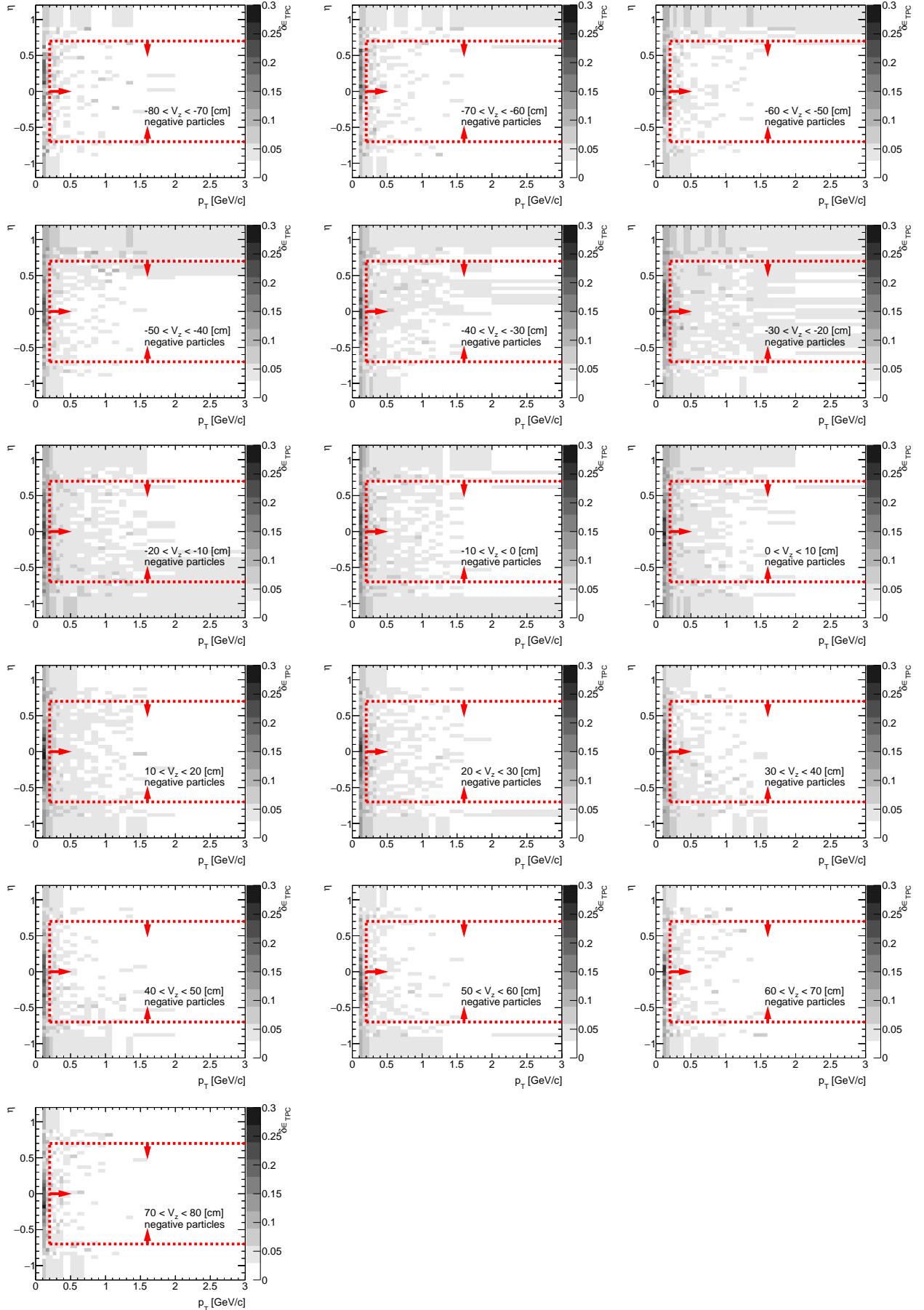


Figure E.15: The amount of lost positive particles due to the interaction with dead material in front of TPC in SD MC sample. Each plot represents the fraction of lost positive particles, $\delta\epsilon_{TPC}$ (z-axis), as a function of true particle pseudorapidity η (y-axis) and transverse momentum p_T (x-axis) in single z -vertex bin.

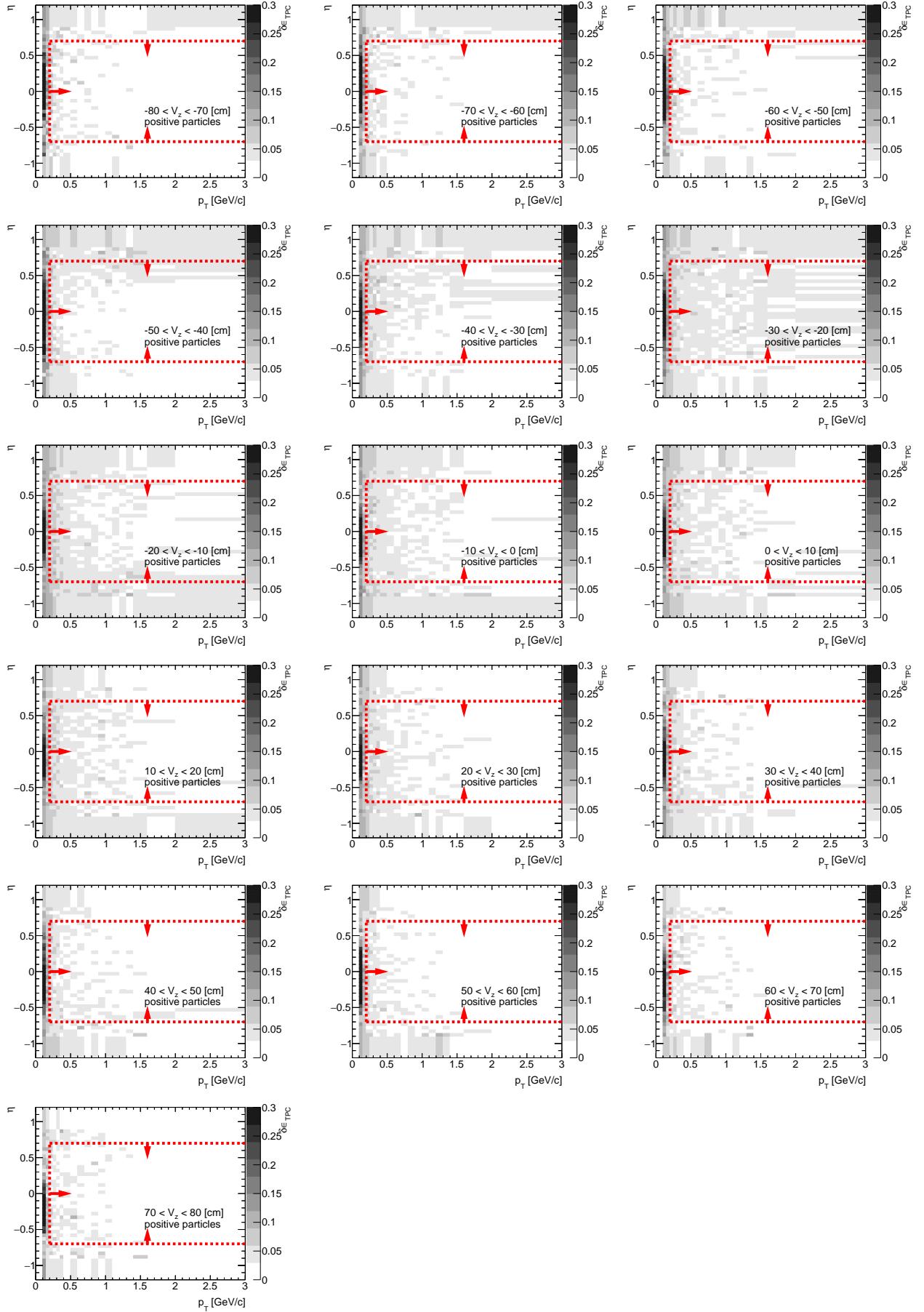
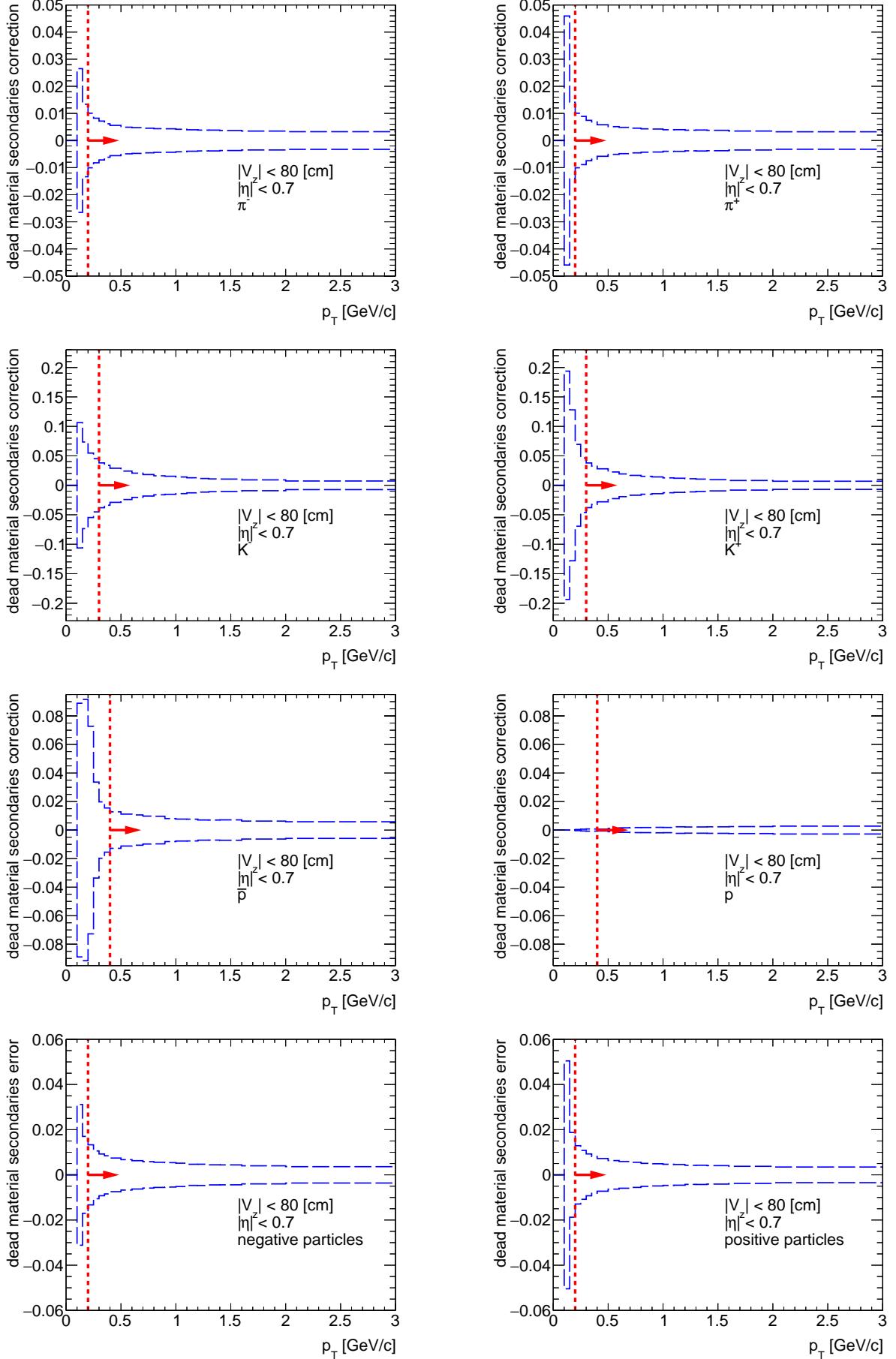


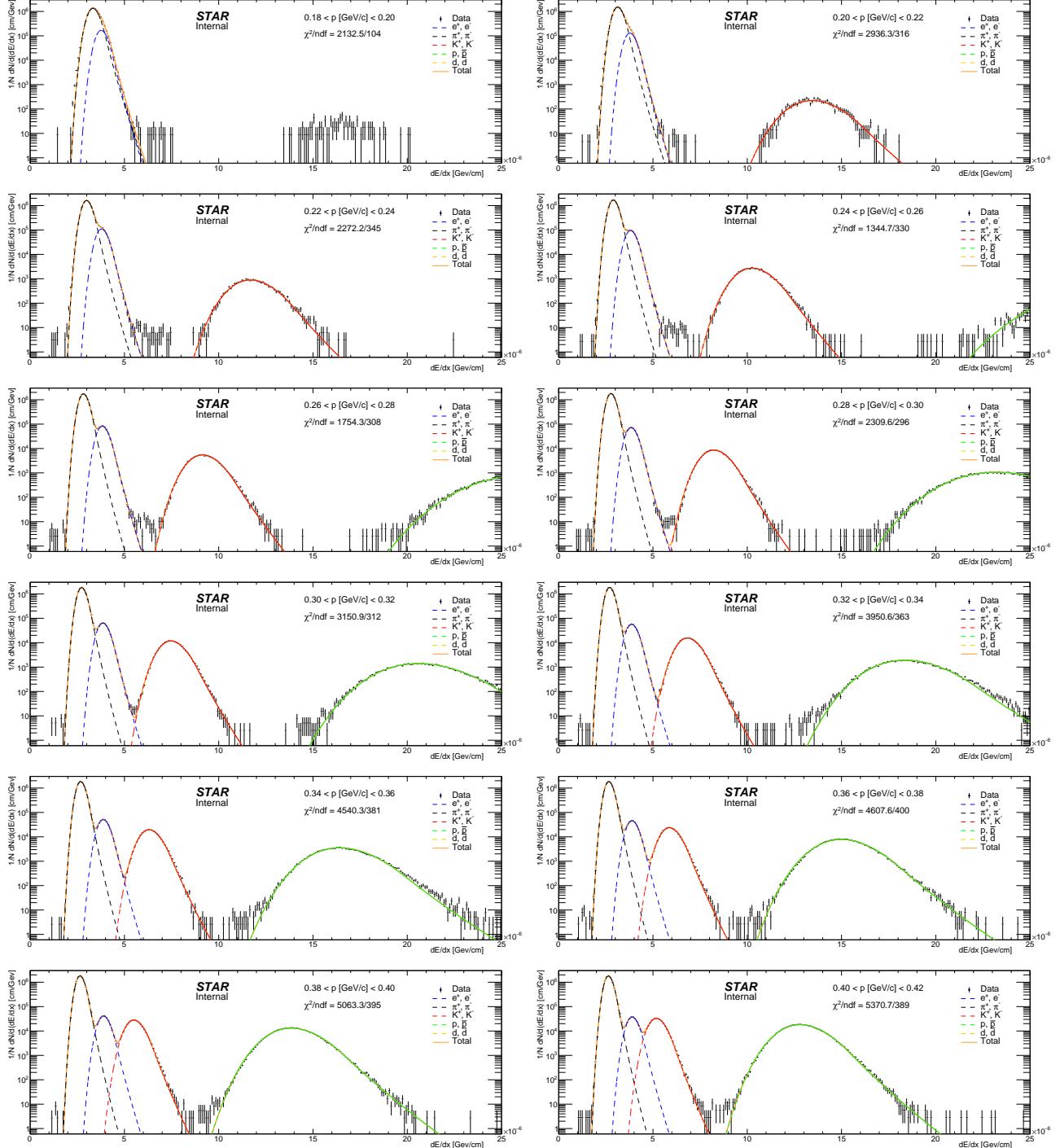
Figure E.16: The systematic uncertainty to the TPC track reconstruction efficiency due to amount of dead material in front of TPC using MC samples for SD. Each plot represents the systematic uncertainty as a function of true particle p_T ($|\eta| < 0.7, |V_z| < 80 \text{ cm}$) for given particle species: $\pi^-, \pi^+, K^-, K^+, \bar{p}$ and p . It was also calculated for negative and positive particles without identification.

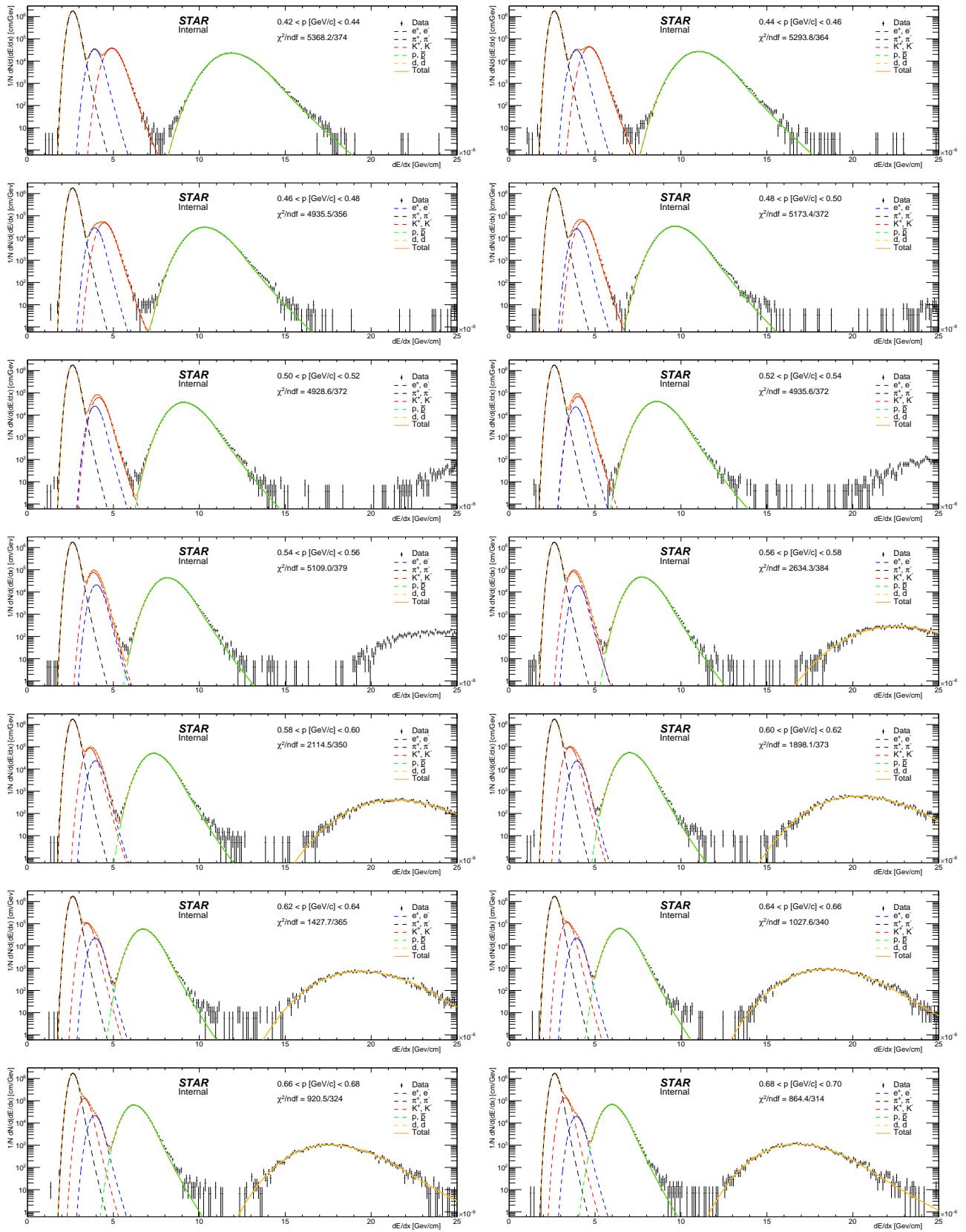


Appendix F

Fits to dE/dx spectra, comparison of dE/dx and n_X^σ between data and MC

Figure F.1: Fits of sum of functions from Eq. (7.6) corresponding to different particle species to dE/dx spectra in the data in momentum bins.





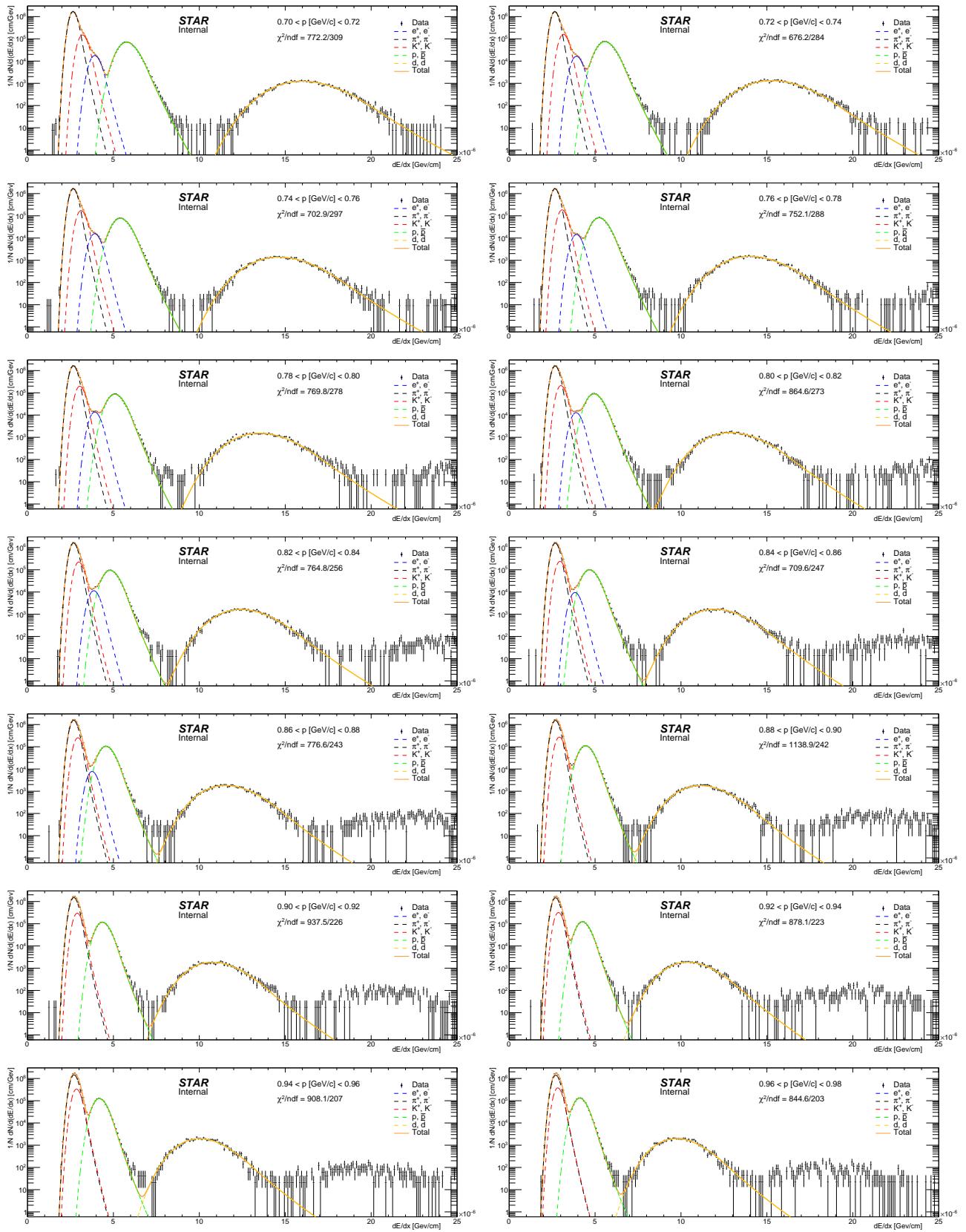
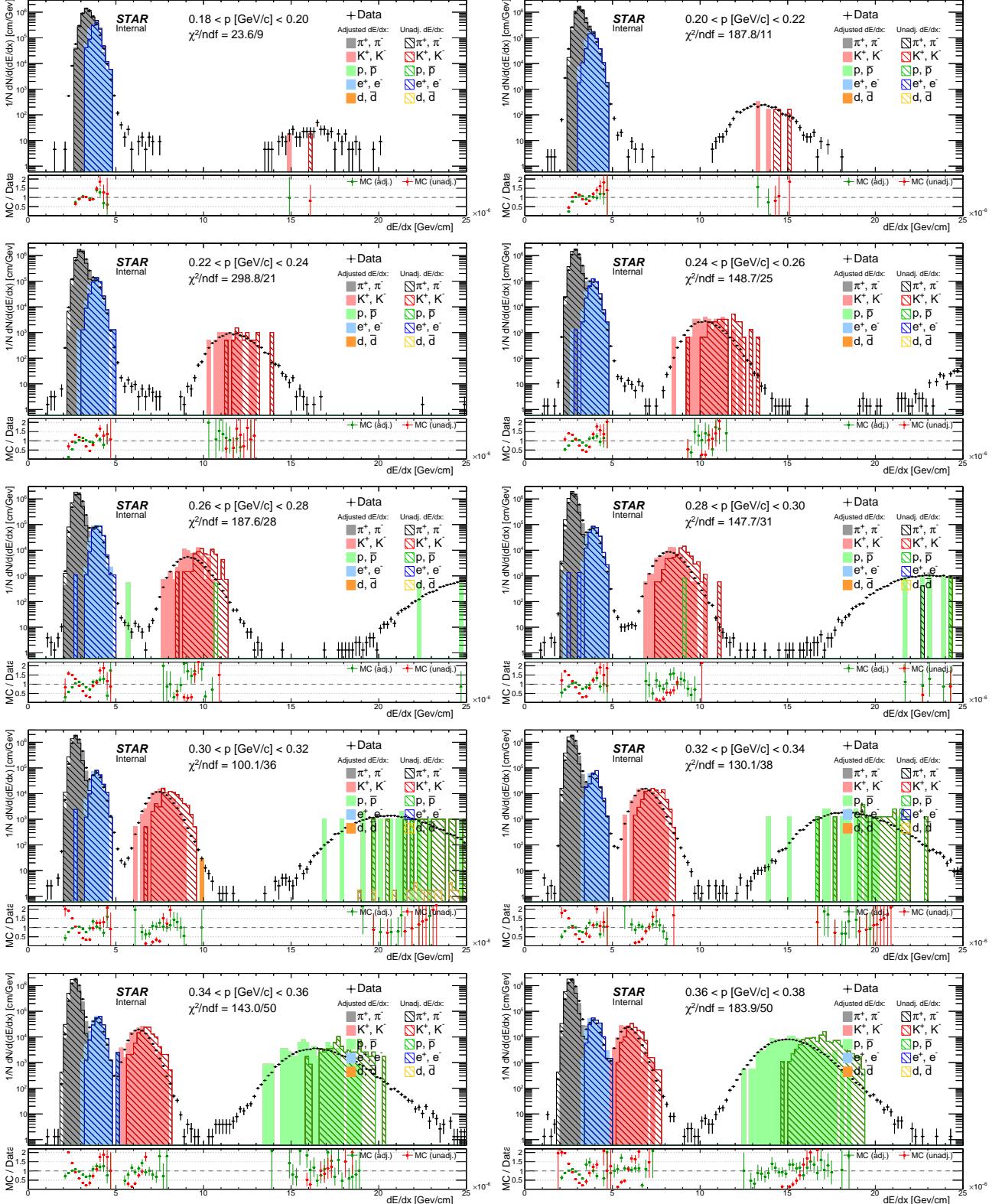
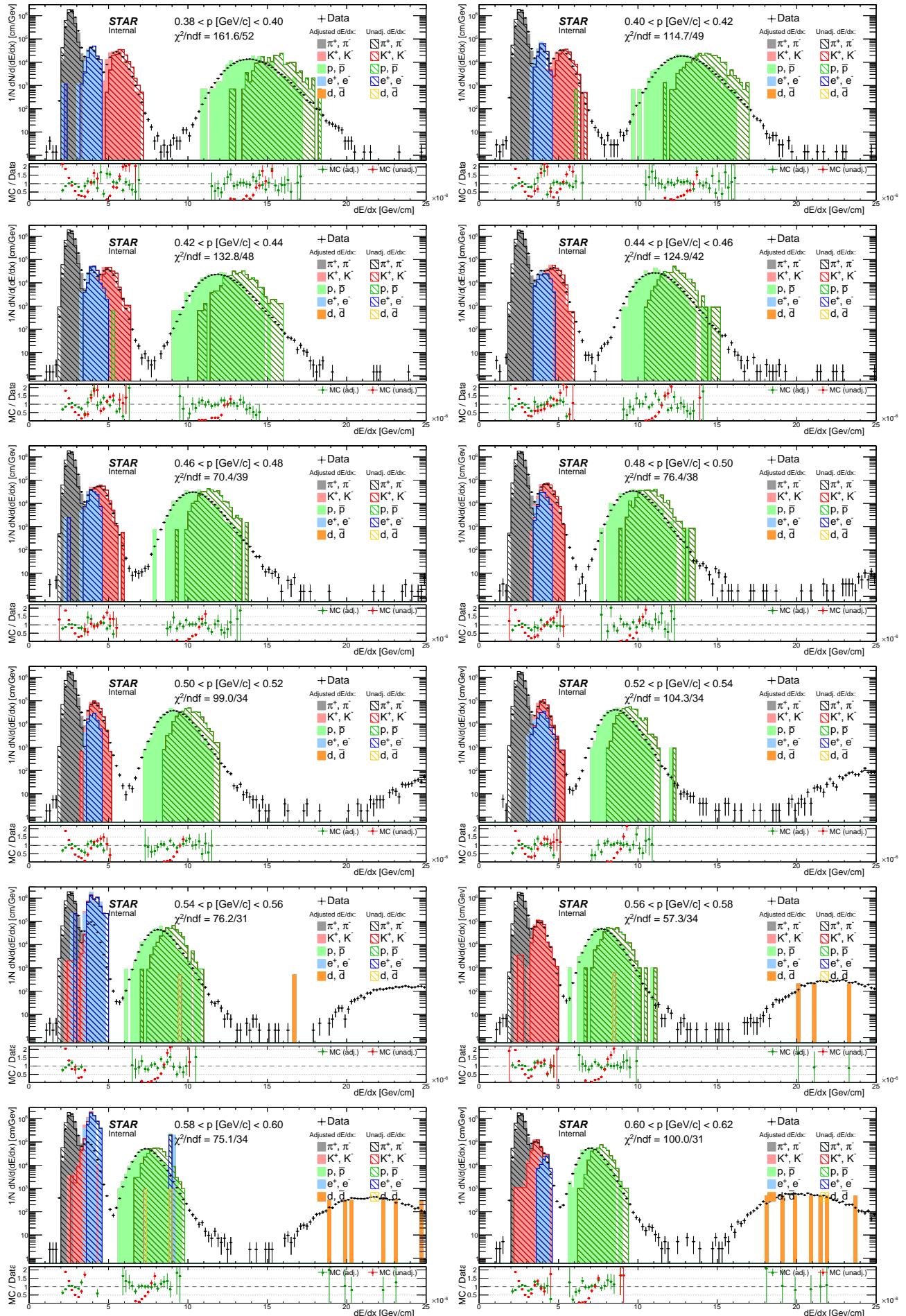
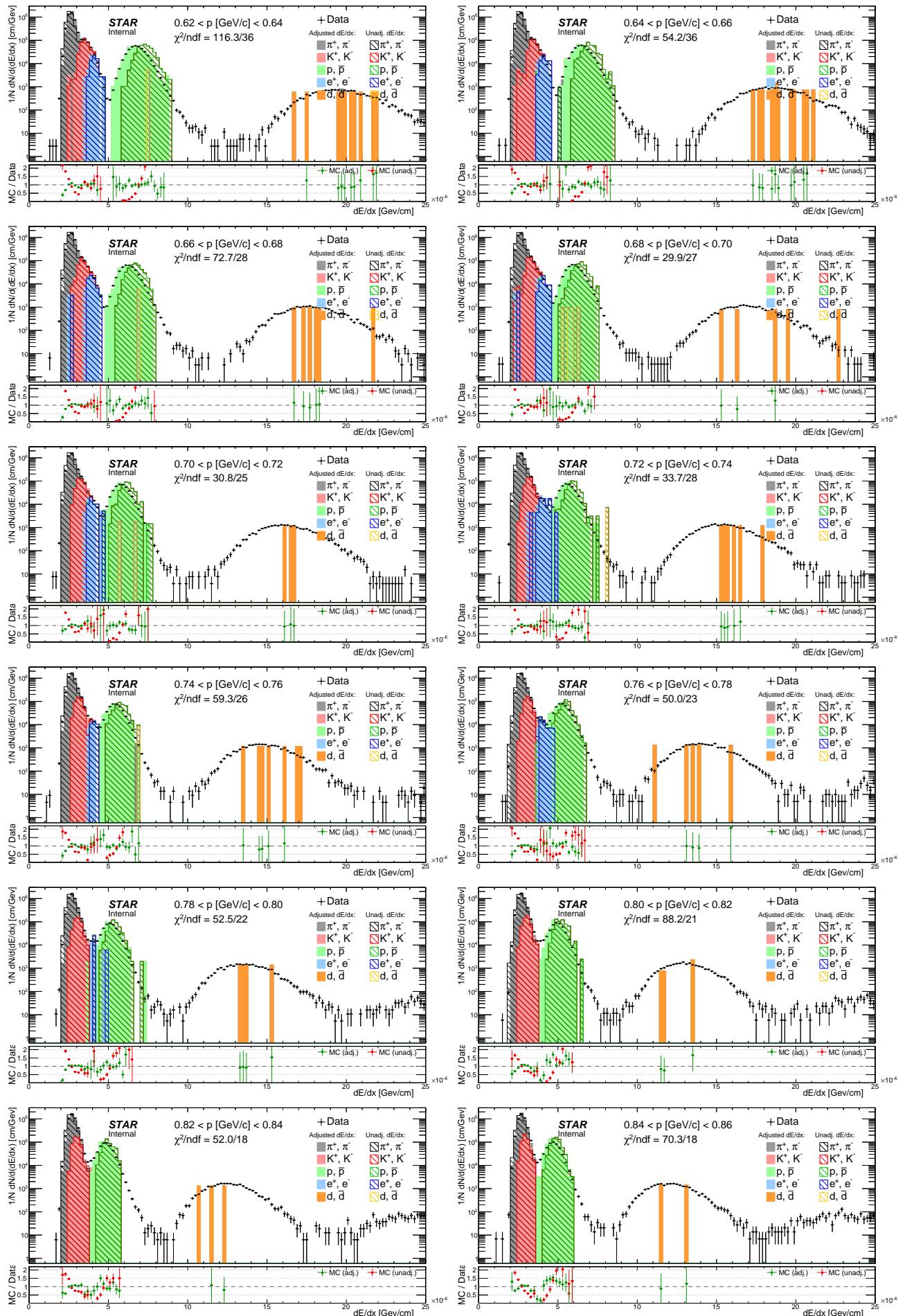


Figure F.2: Comparison of dE/dx spectrum between data and embedded MC (before and after dE/dx adjustment) in momentum bins.







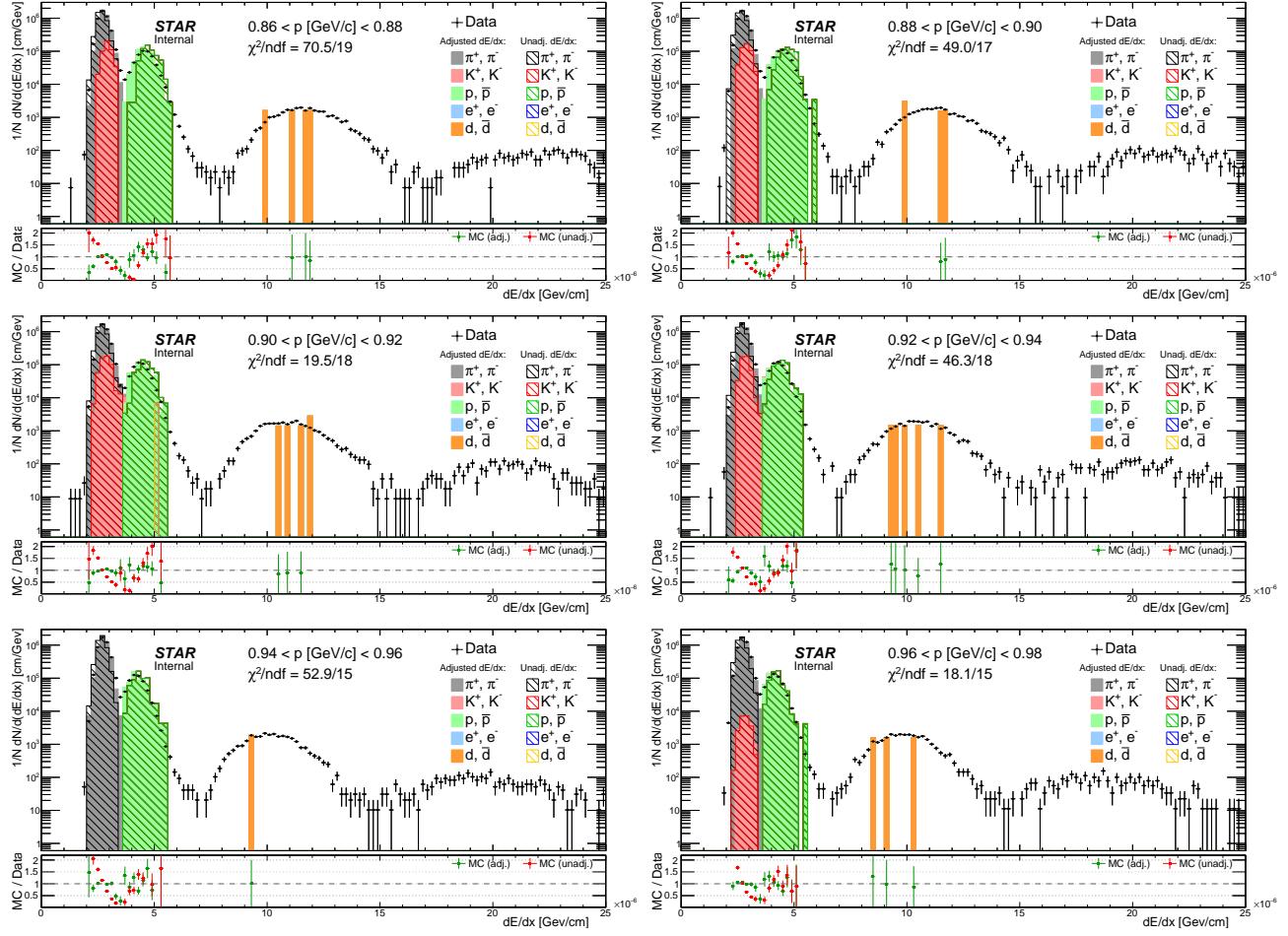
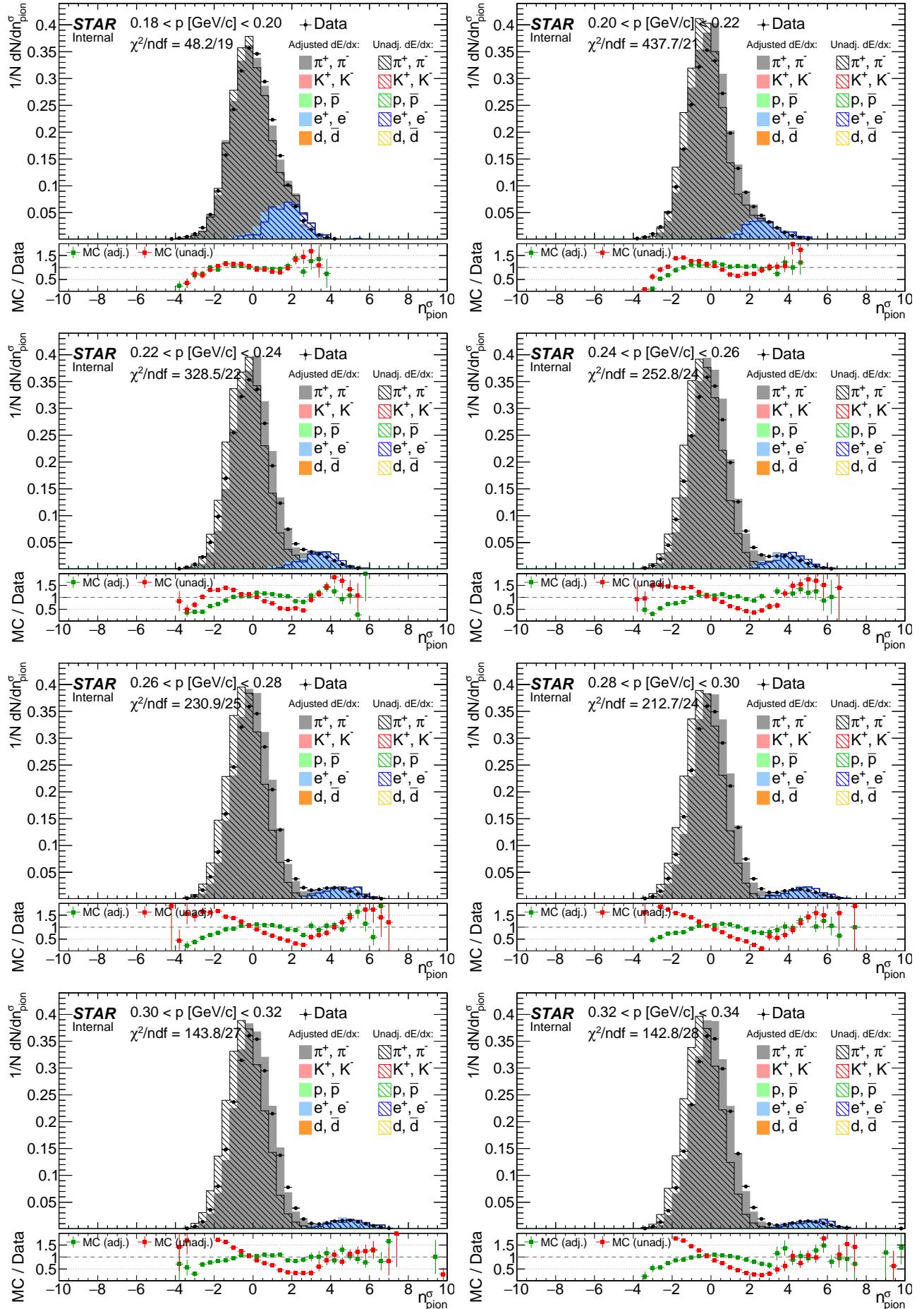
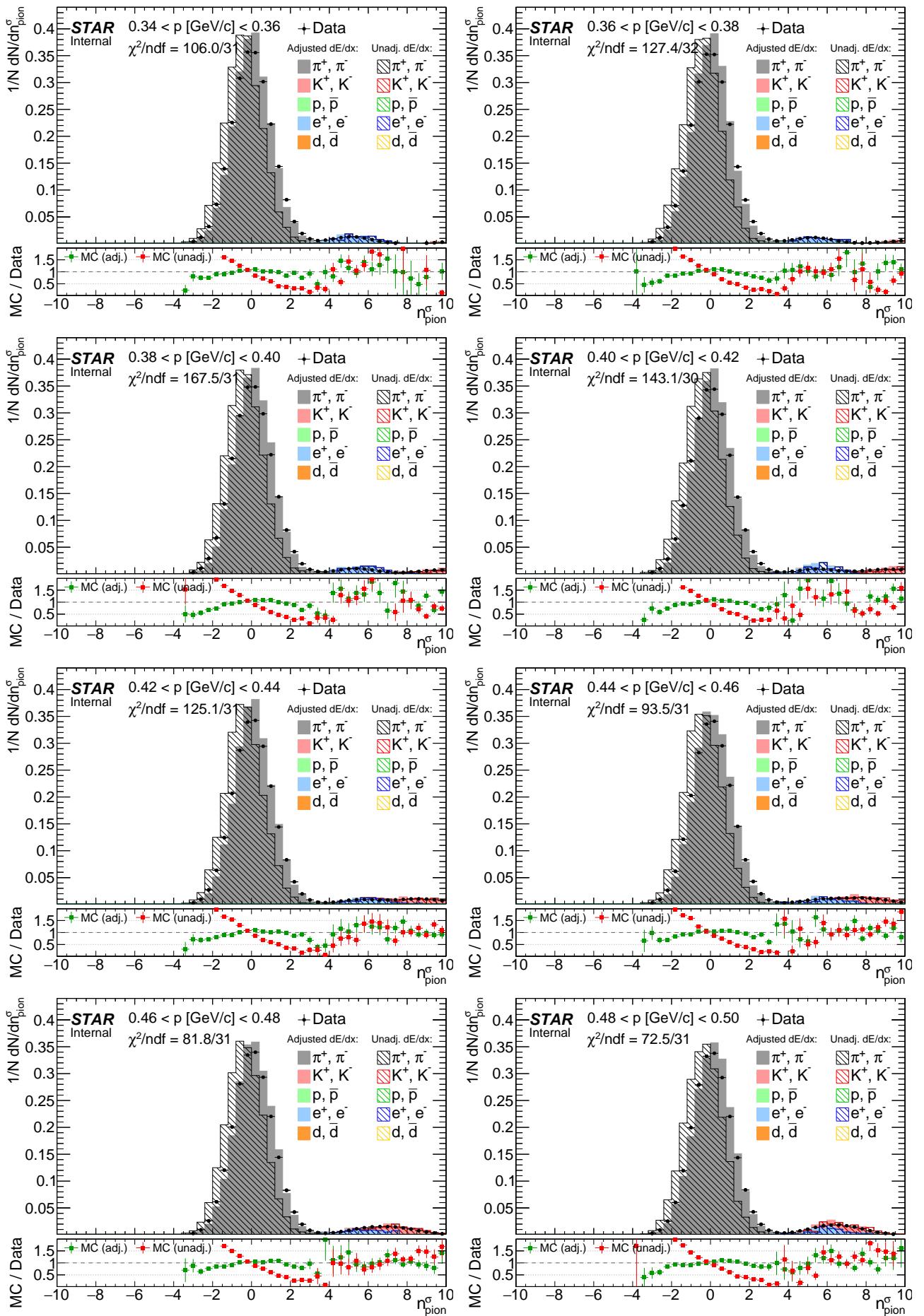
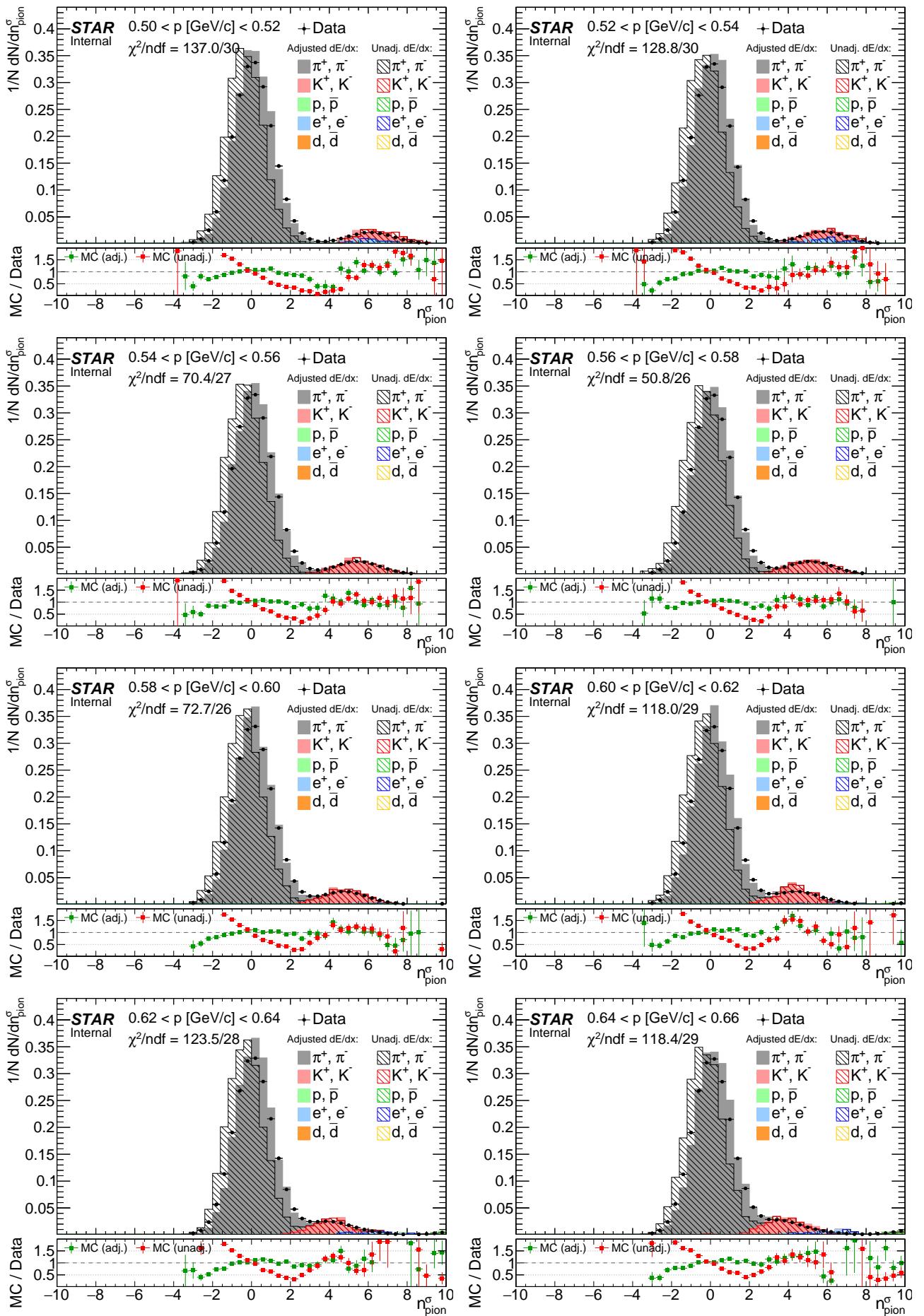
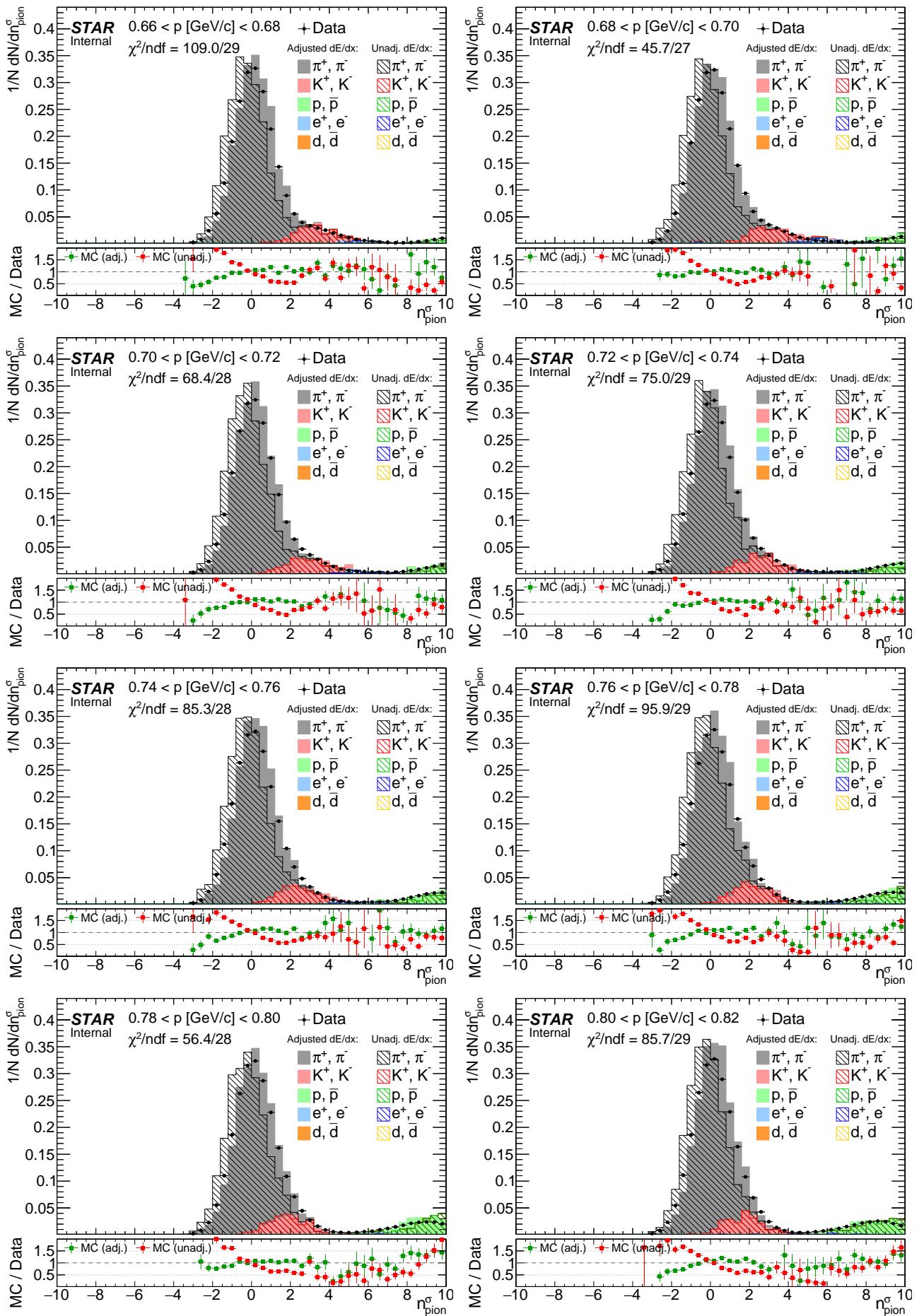


Figure F.3: Comparison of n_{pion}^{σ} distribution between data and embedded MC (before and after dE/dx adjustment) in momentum bins.









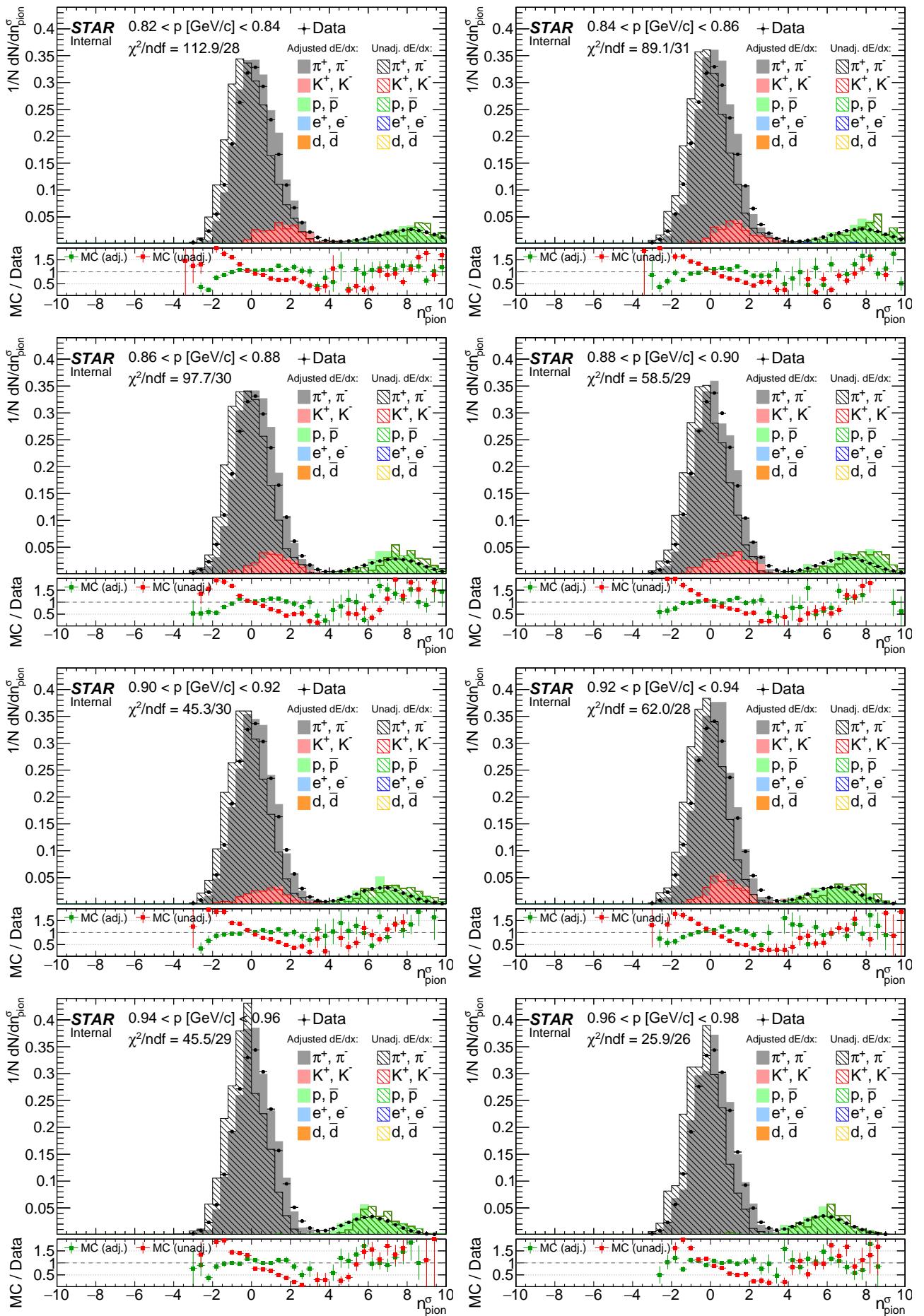
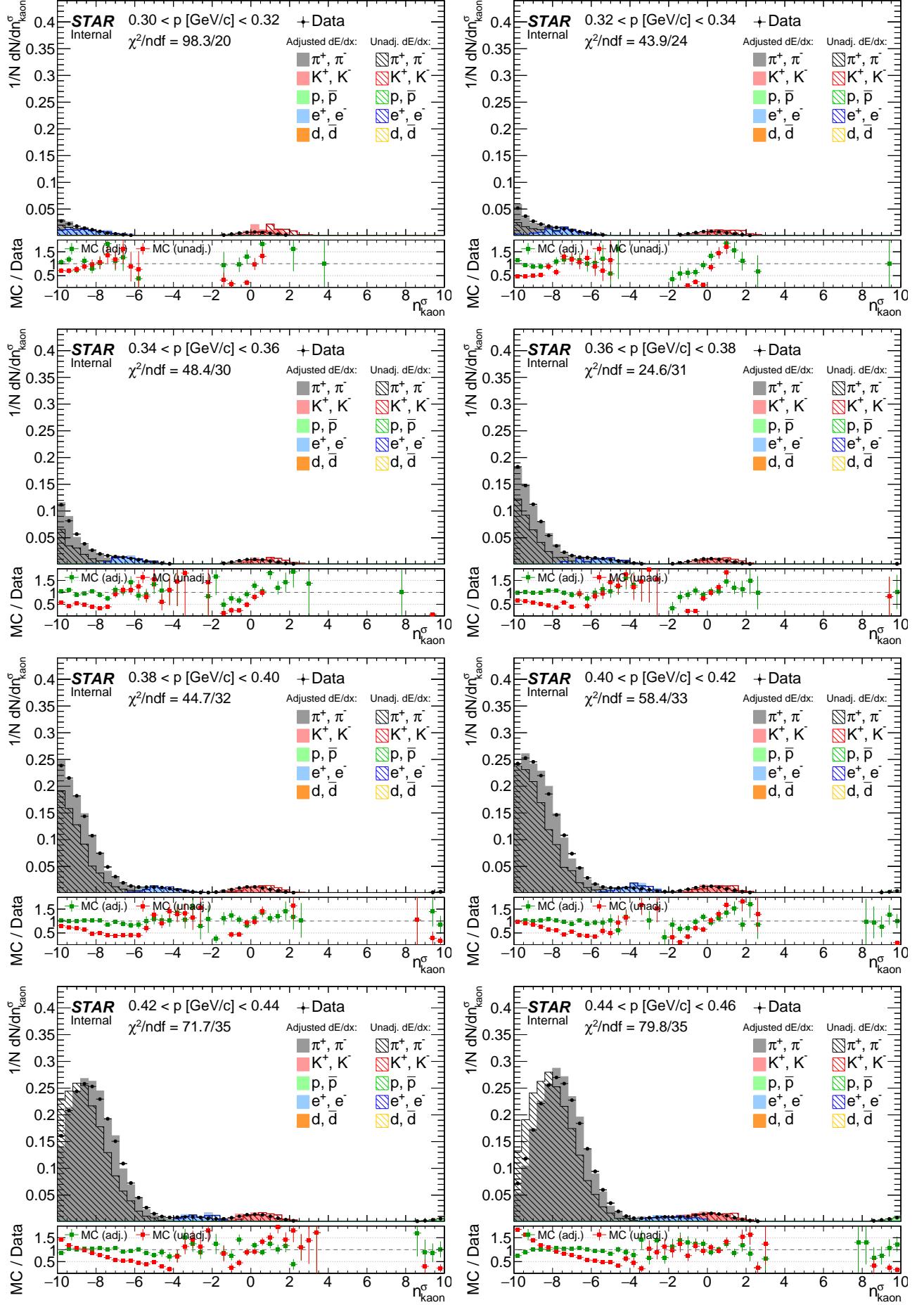
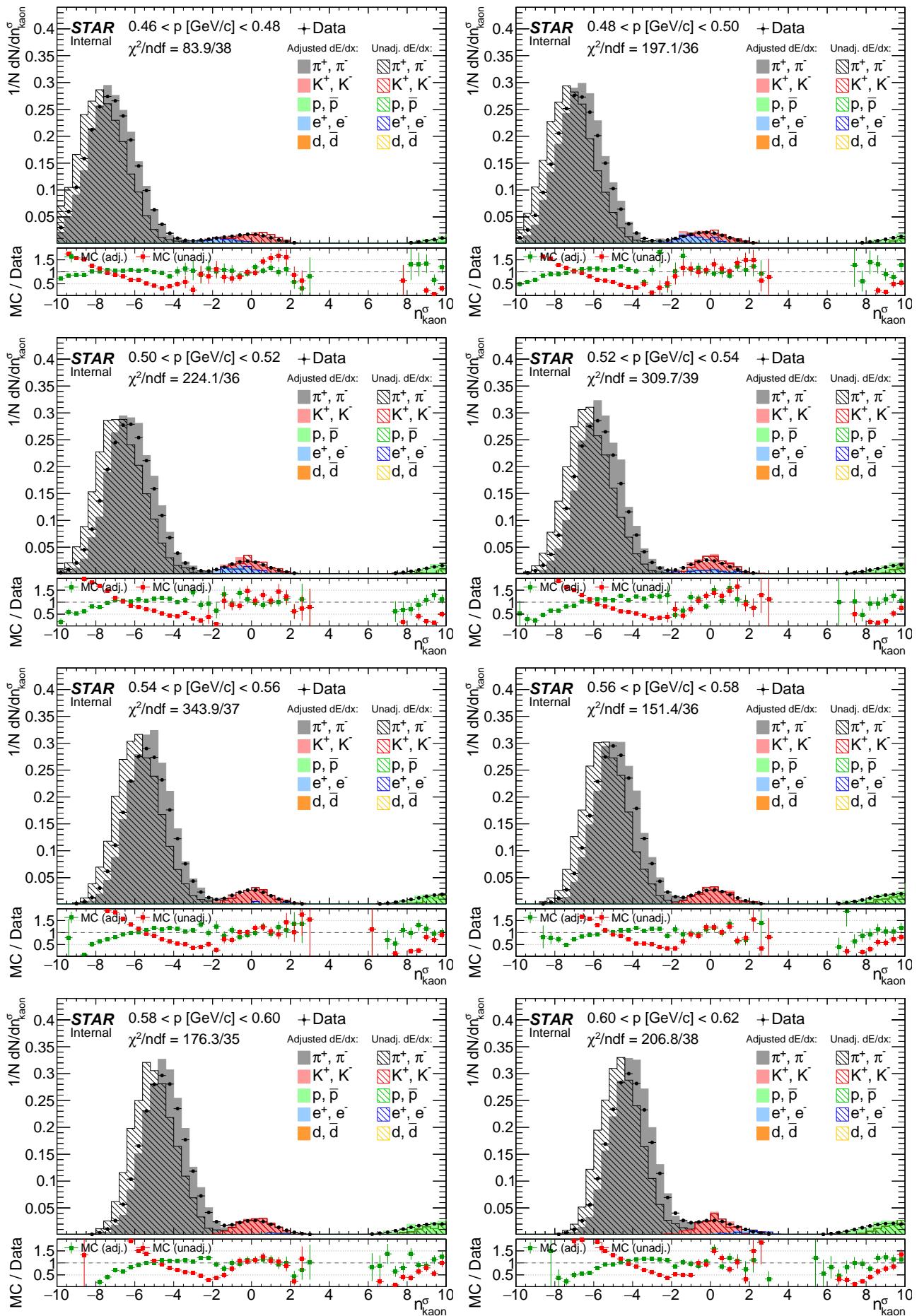
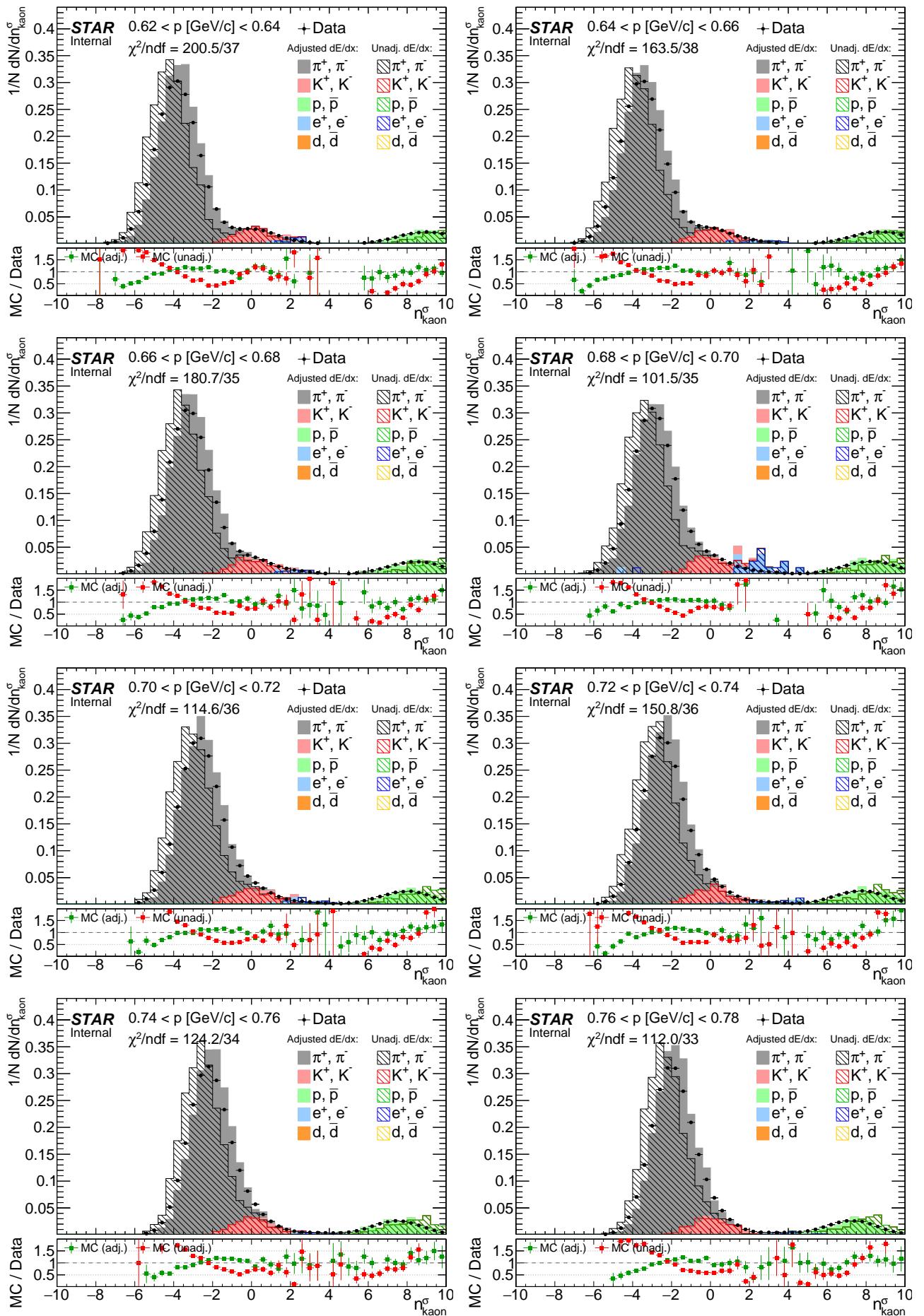
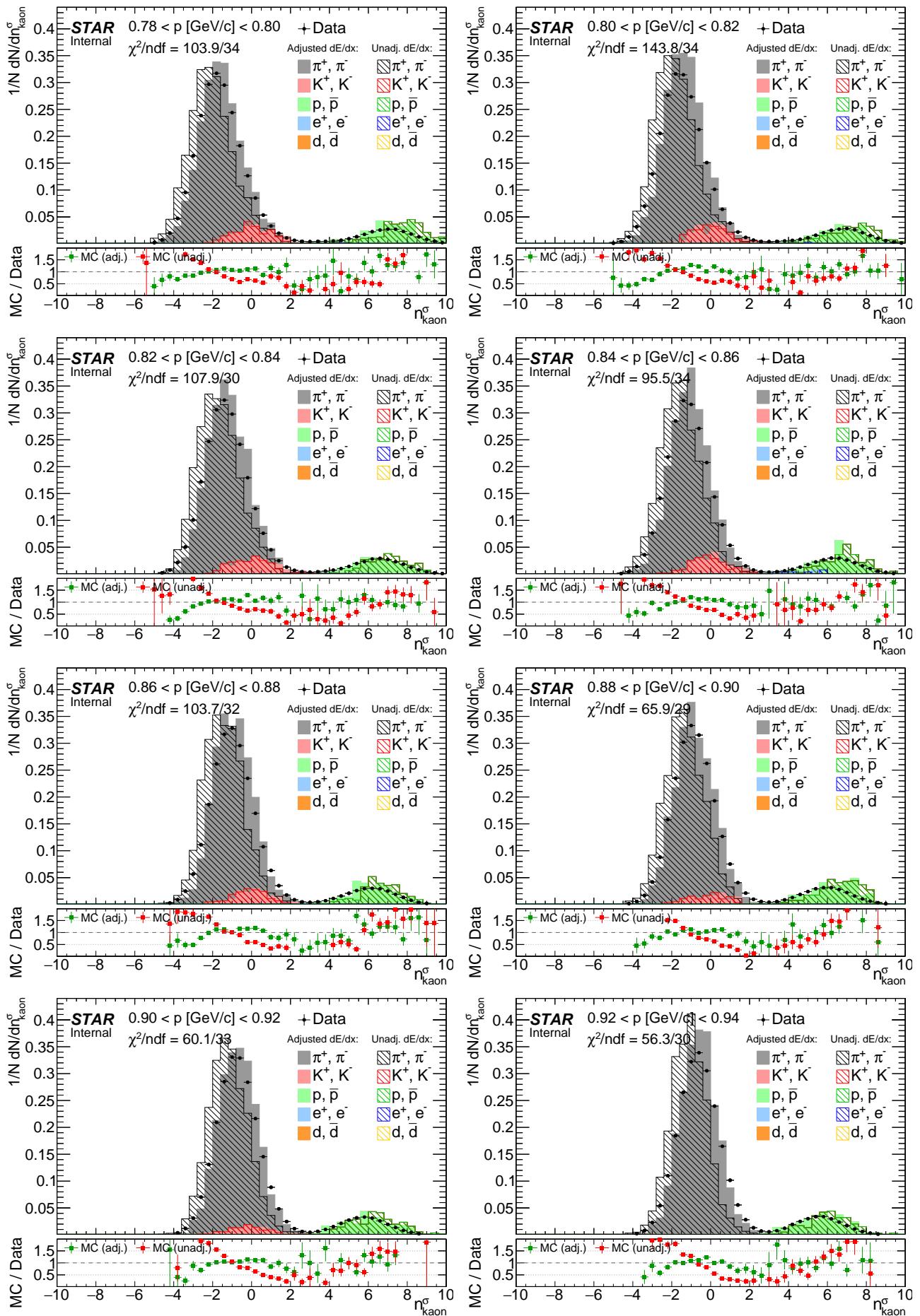


Figure F.4: Comparison of n_{kaon}^{σ} distribution between data and embedded MC (before and after dE/dx adjustment) in momentum bins.









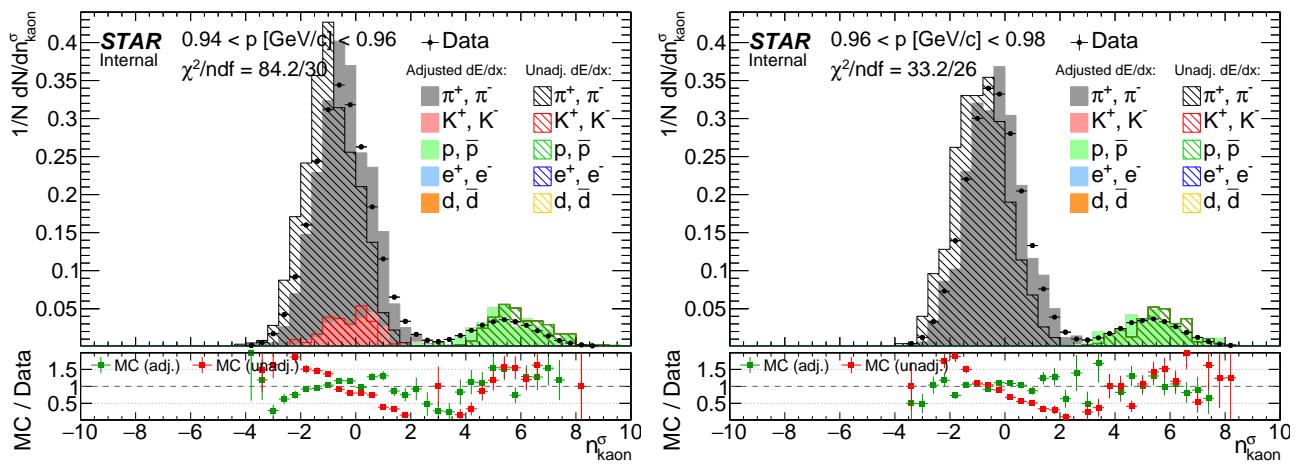
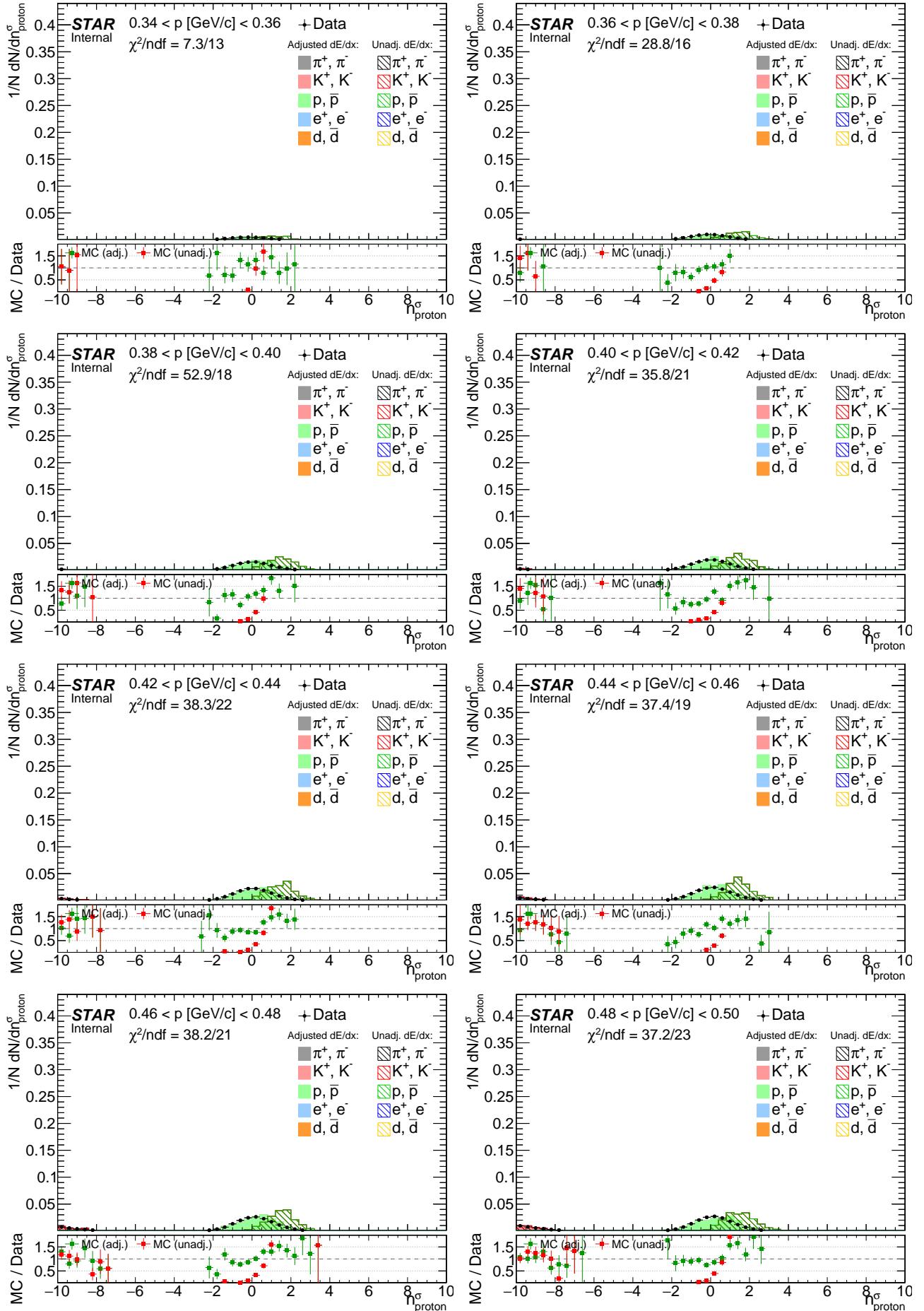
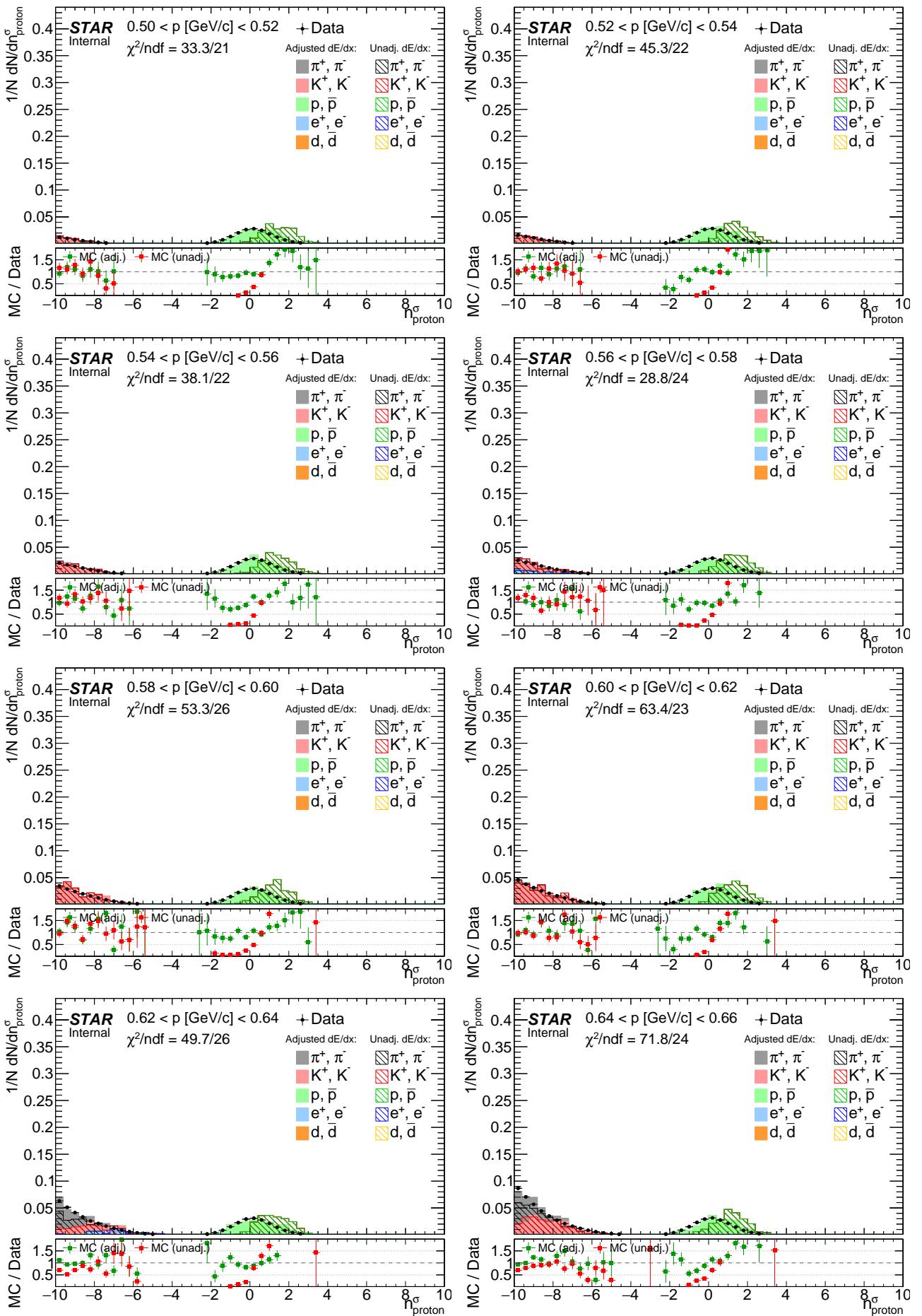
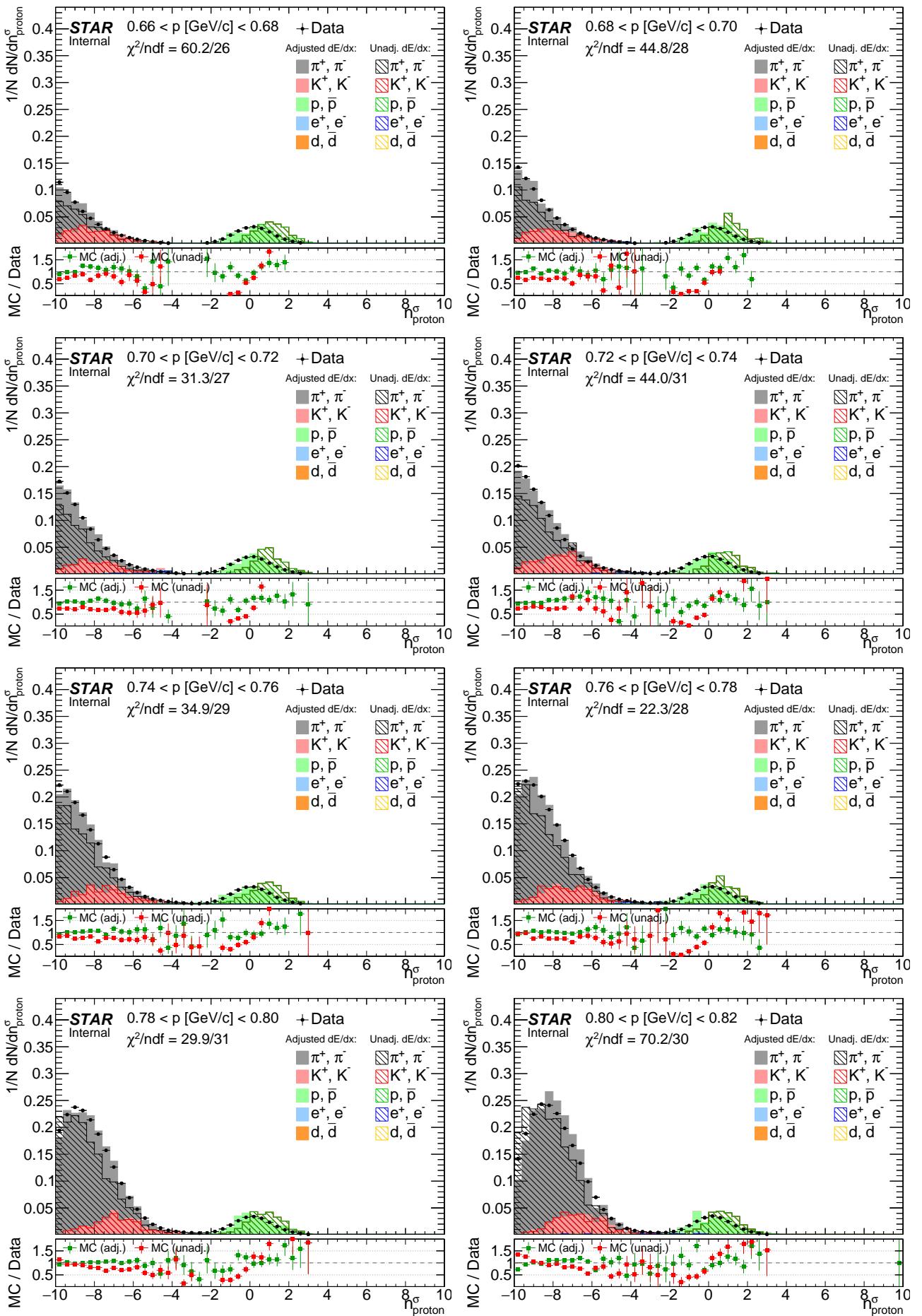
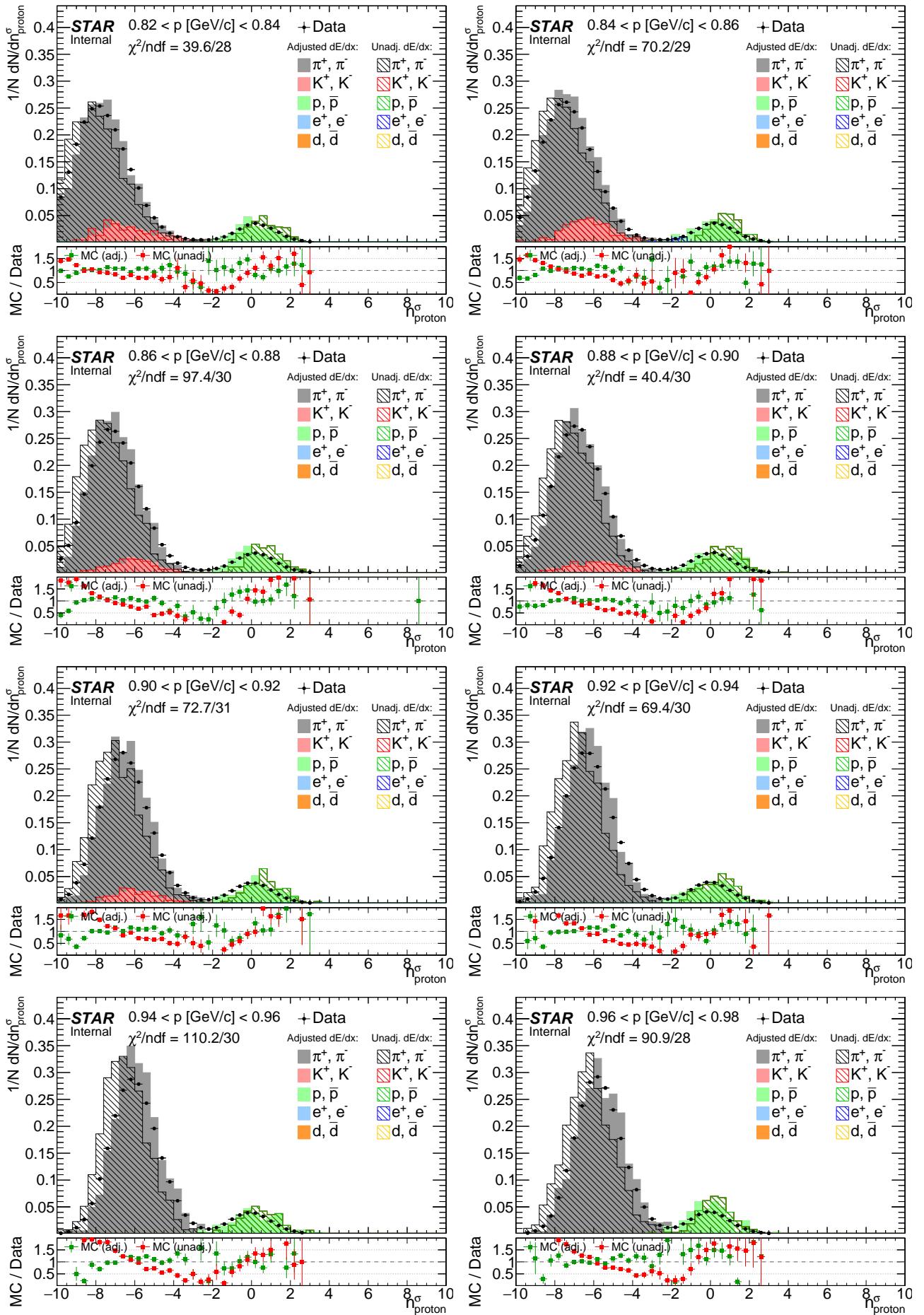


Figure F.5: Comparison of $n_{\text{proton}}^{\sigma}$ distribution between data and embedded MC (before and after dE/dx adjustment) in momentum bins.









Appendix G

Tag&Probe fits for TOF hit reconstruction and matching efficiency

Figure G.1: Total transverse momentum p_T^{miss} of the $p+\text{Tag}+\text{Probe}+p$ system in the data and signal+background embedded MC, in bins of p_T of a probe. Adjacent plots are for the same p_T bin, one for data (left) and the other for MC (right).

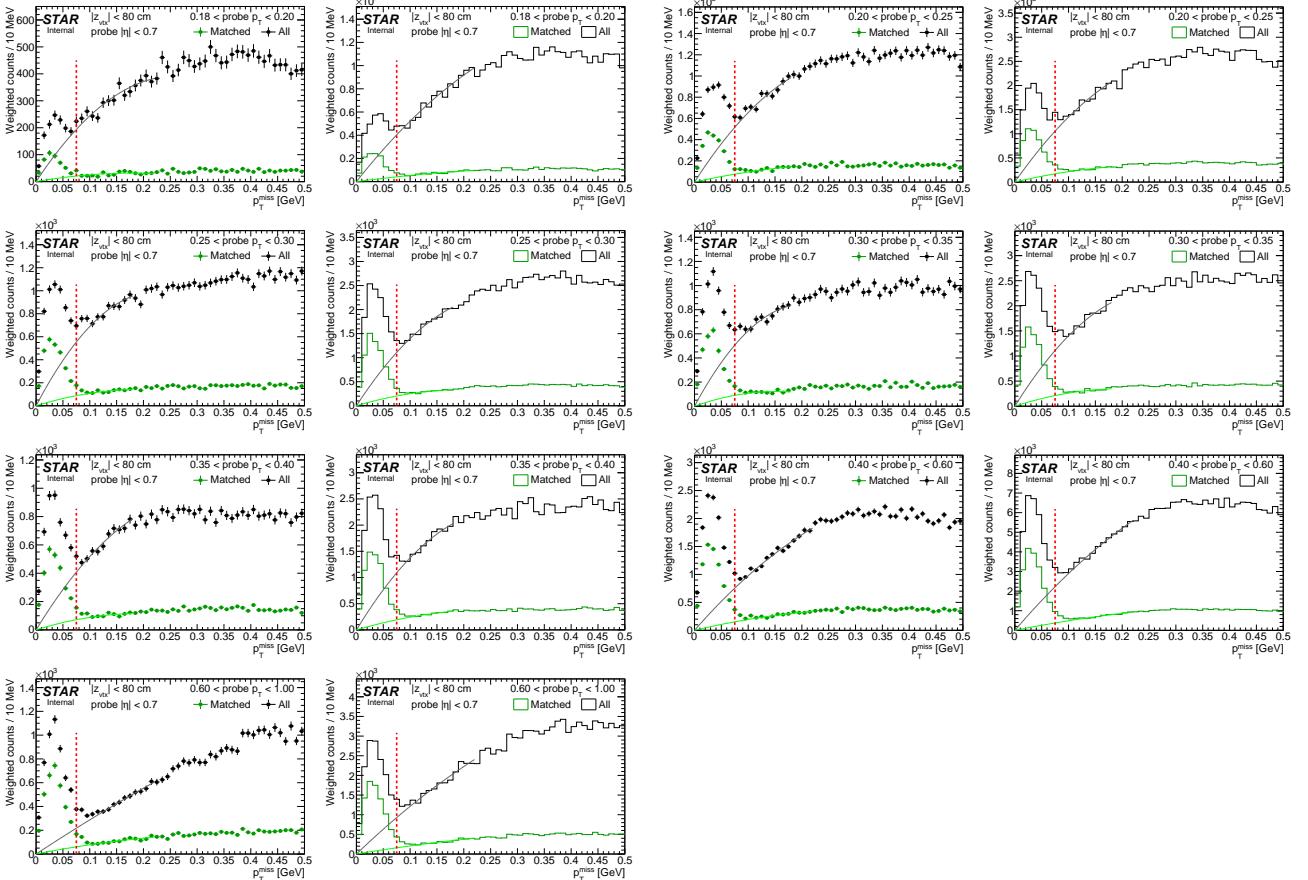
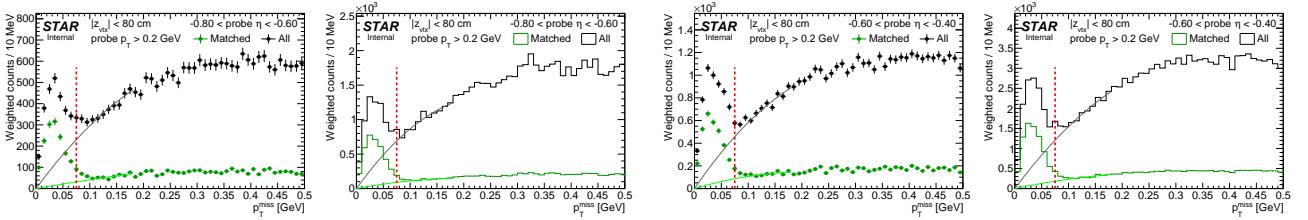


Figure G.2: Total transverse momentum p_T^{miss} of the $p+\text{Tag}+\text{Probe}+p$ system in the data and signal+background embedded MC, in bins of η of a probe. Adjacent plots are for the same η bin, one for data (left) and the other for MC (right).



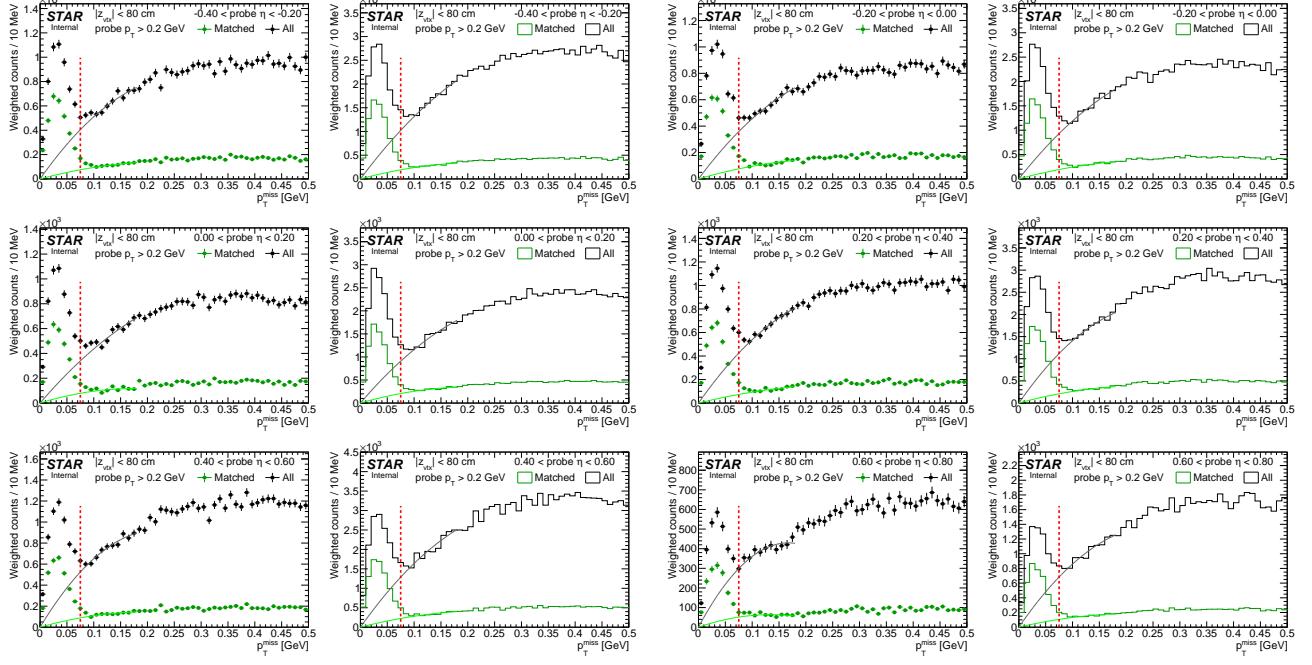


Figure G.3: Total transverse momentum p_T^{miss} of the $p+\text{Tag}+\text{Probe}+p$ system in the data and signal+background embedded MC, in bins of z_{vtx} of a probe. Adjacent plots are for the same z_{vtx} bin, one for data (left) and the other for MC (right).

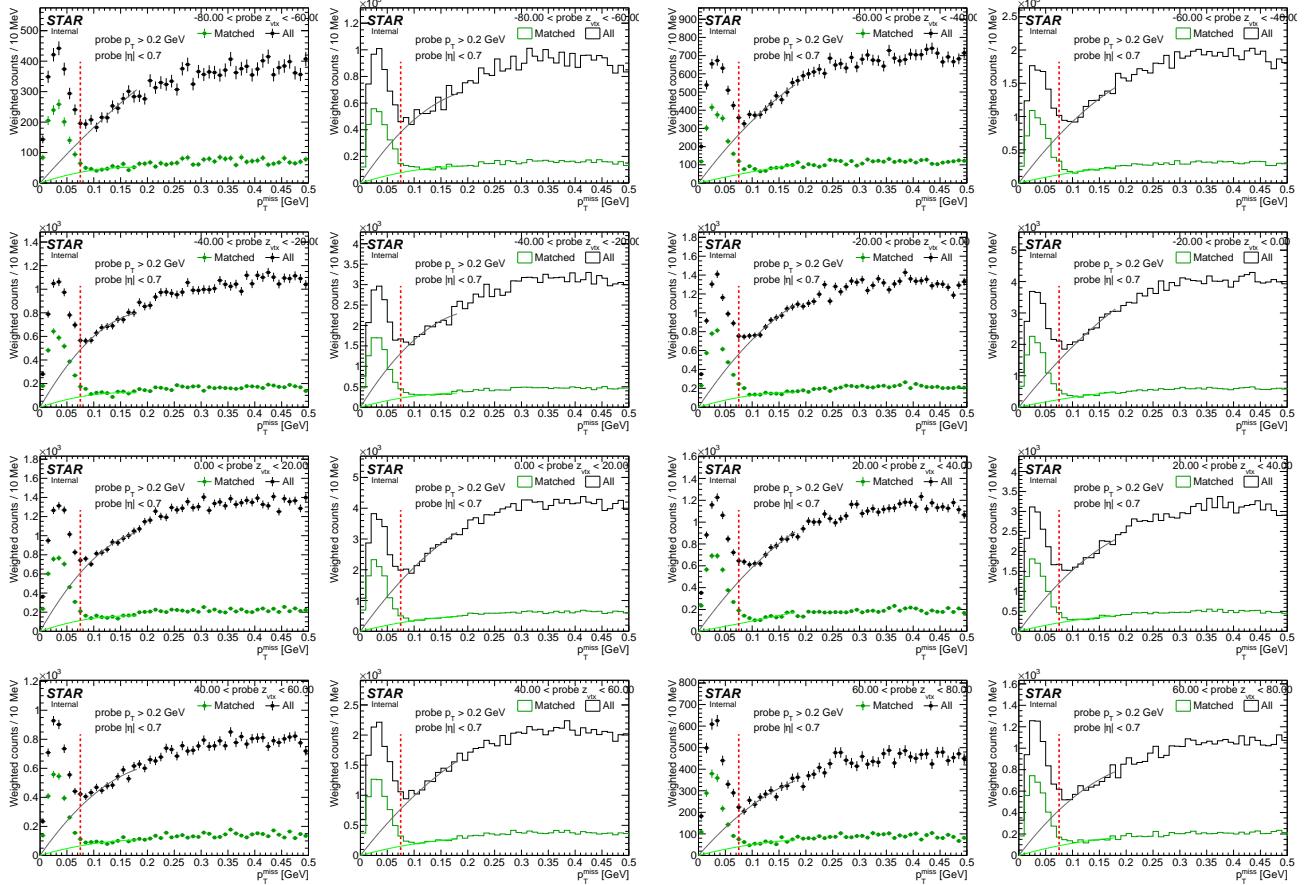


Figure G.4: Total transverse momentum p_T^{miss} of the p +Tag+Probe+ p system in the data and signal+background embedded MC, in bins of p_T of a probe for probe pseudorapidity range $-0.7 < \eta < -0.3$. Adjacent plots are for the same p_T bin, one for data (left) and the other for MC (right).

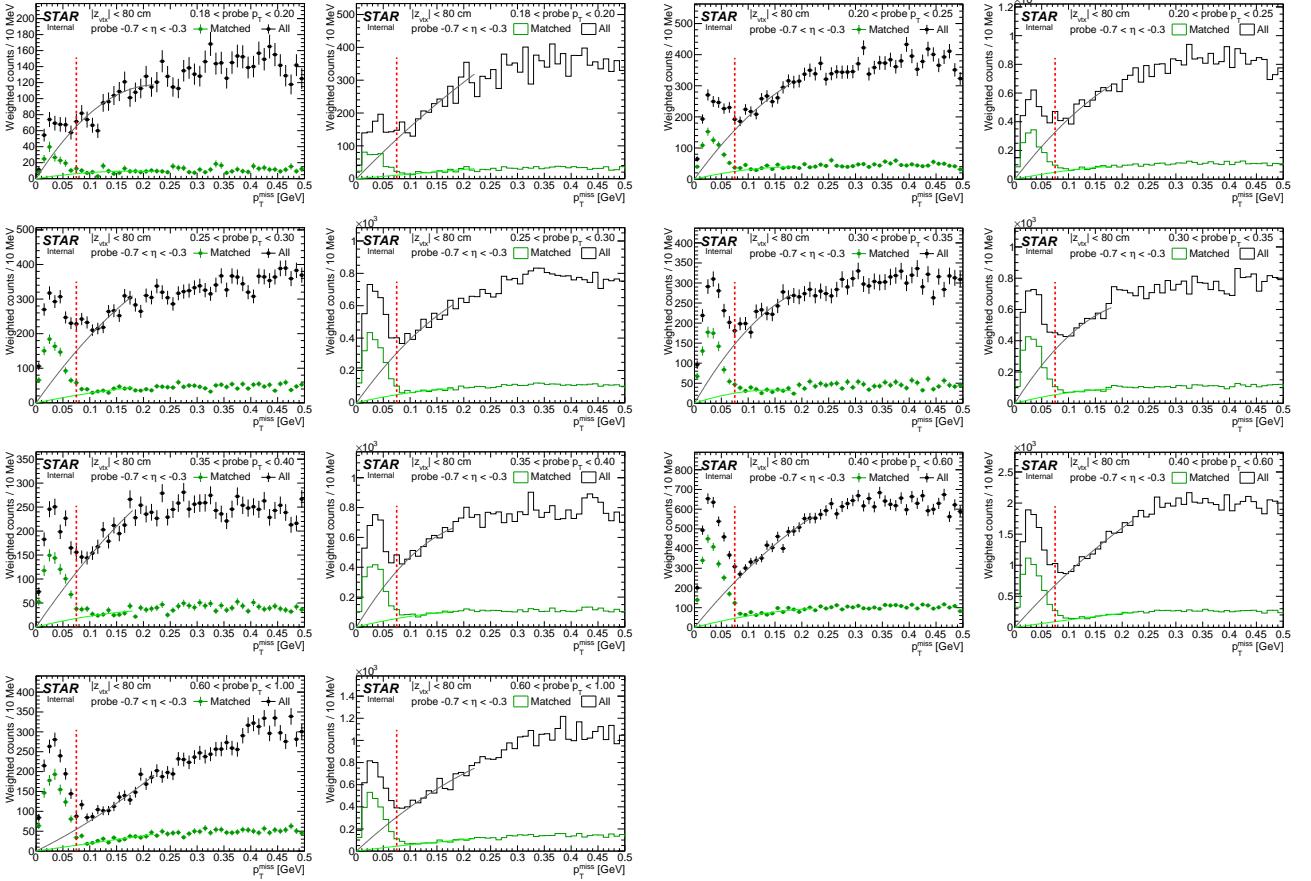
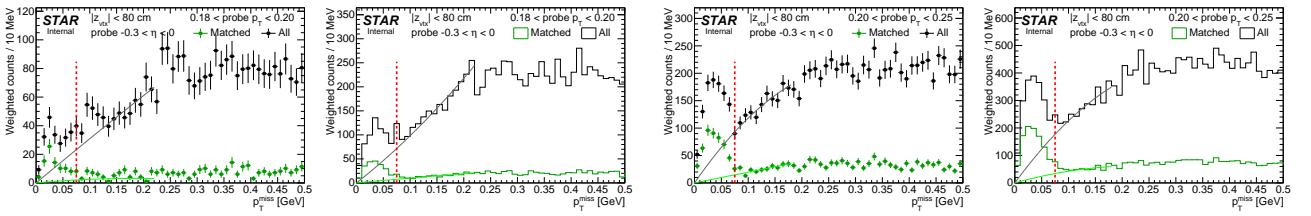


Figure G.5: Total transverse momentum p_T^{miss} of the p +Tag+Probe+ p system in the data and signal+background embedded MC, in bins of p_T of a probe for probe pseudorapidity range $-0.3 < \eta < 0$. Adjacent plots are for the same p_T bin, one for data (left) and the other for MC (right).



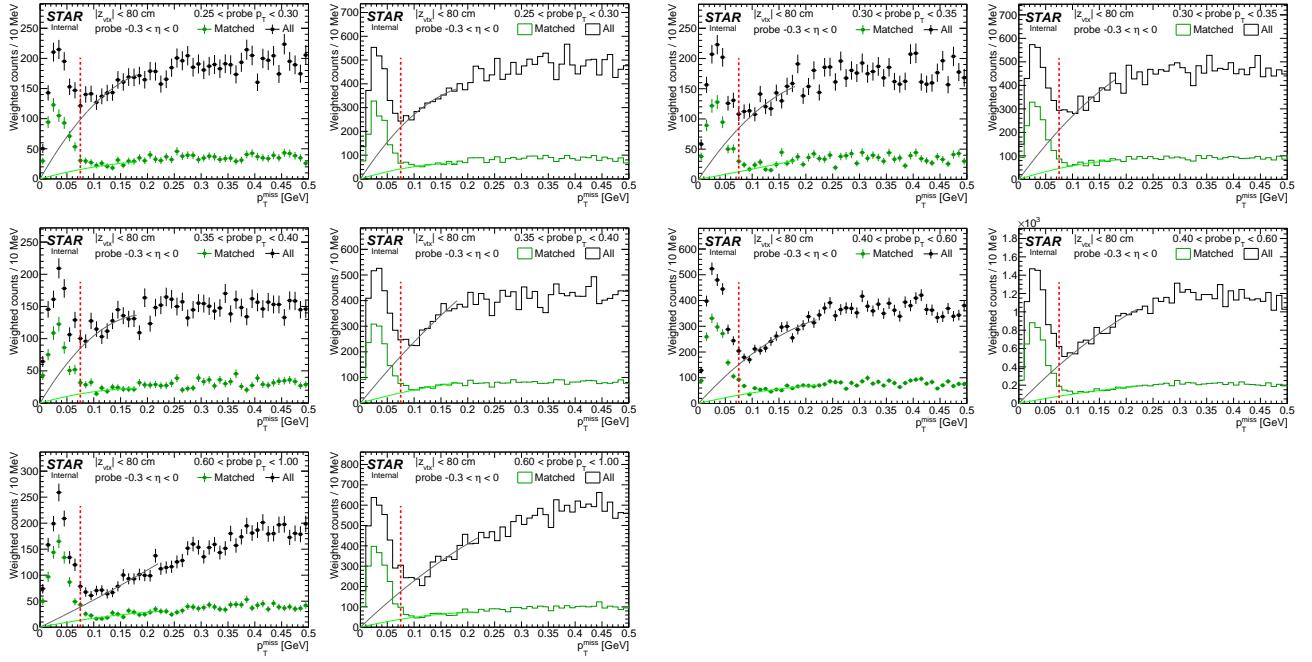


Figure G.6: Total transverse momentum p_T^{miss} of the $p+\text{Tag}+\text{Probe}+p$ system in the data and signal+background embedded MC, in bins of p_T of a probe for probe pseudorapidity range $0 < \eta < 0.3$. Adjacent plots are for the same p_T bin, one for data (left) and the other for MC (right).

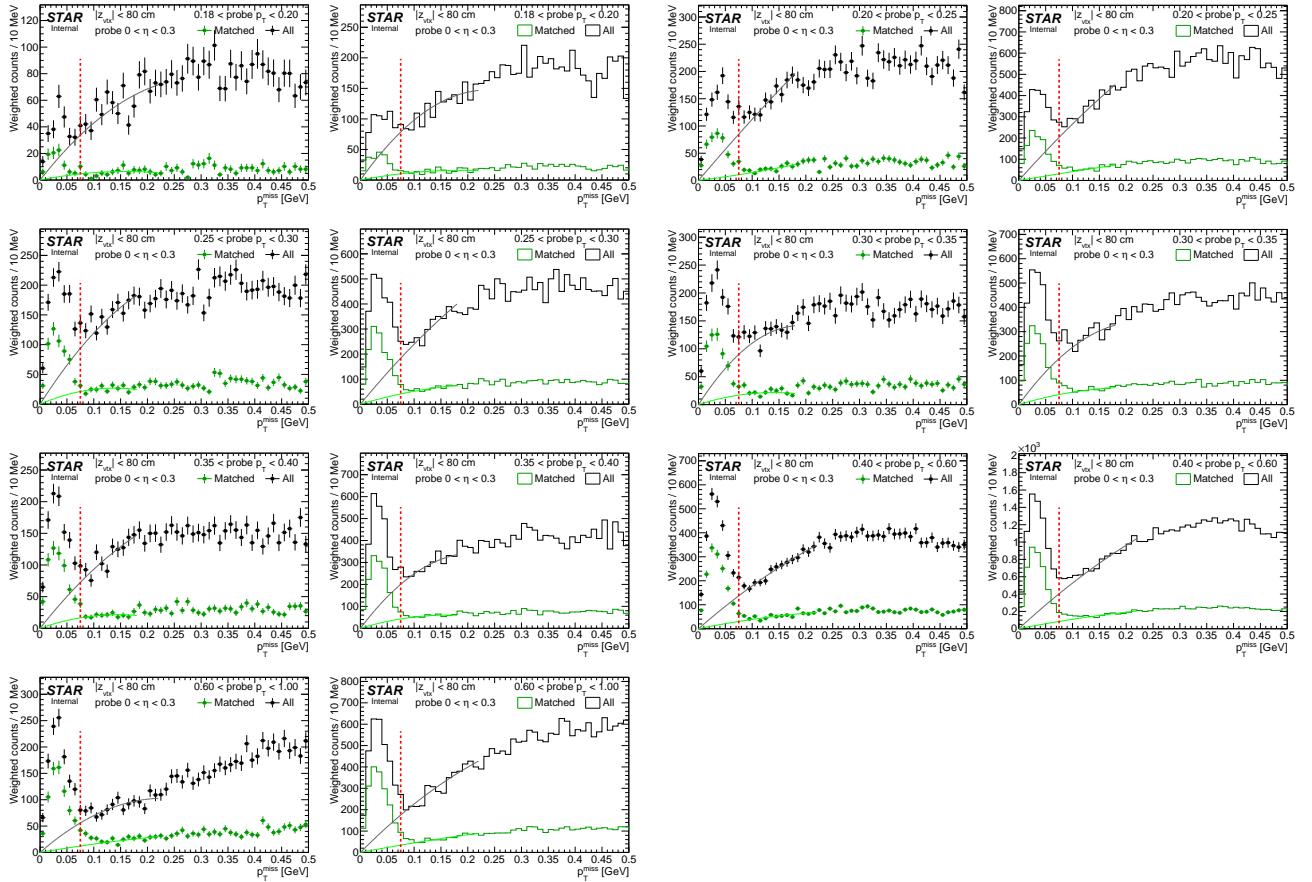
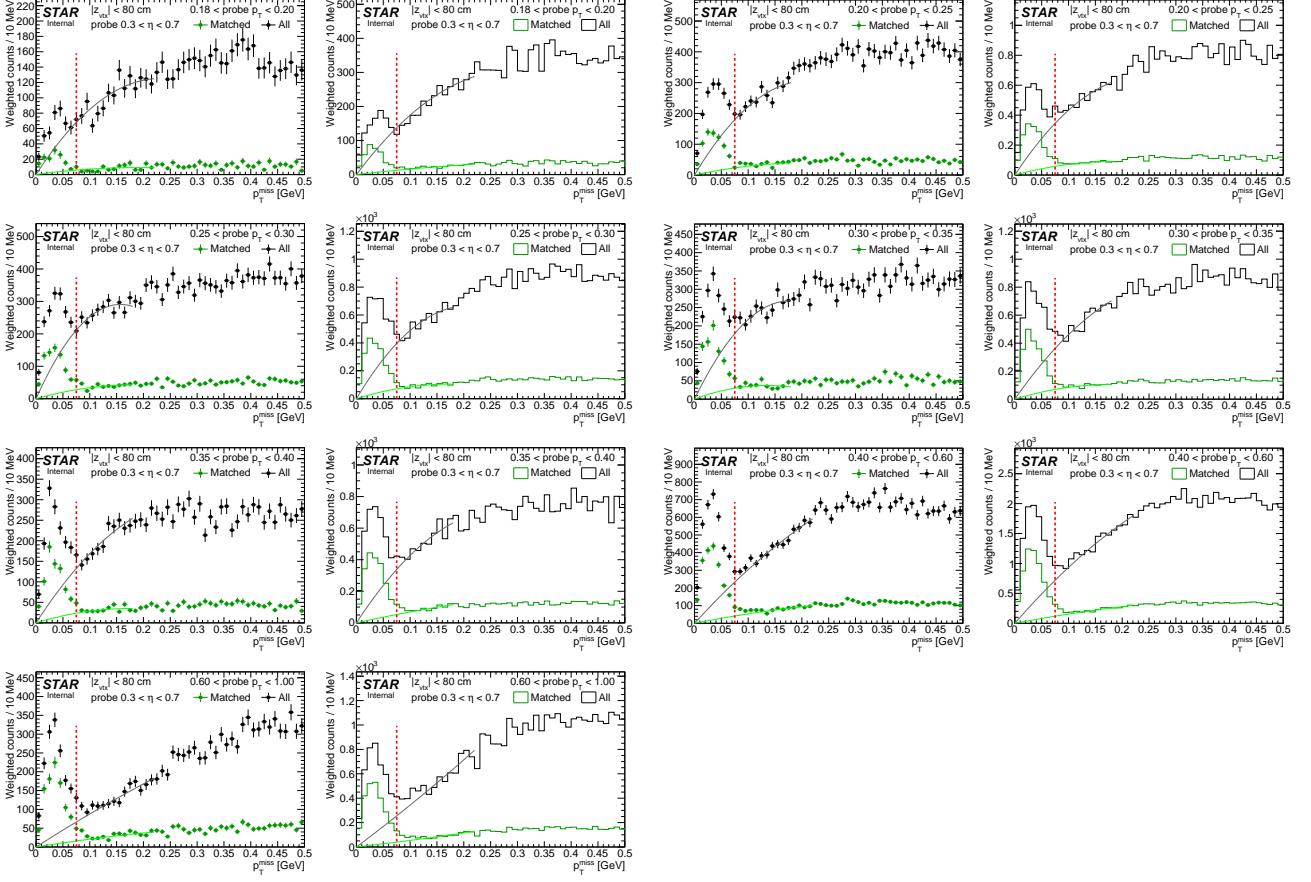


Figure G.7: Total transverse momentum p_T^{miss} of the p +Tag+Probe+ p system in the data and signal+background embedded MC, in bins of p_T of a probe for probe pseudorapidity range $0.3 < \eta < 0.7$. Adjacent plots are for the same p_T bin, one for data (left) and the other for MC (right).



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