

A preliminary analysis of data from ALICE's new ITS and MFT detectors

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Abstract

We investigate pilot-beam data from proton-proton collisions in Run 3 at ALICE, tracing its journey through two of the new detectors (the Inner Tracking System and Muon Forward Tracker) and the new Online-Offline analysis framework. The data is analysed to assess the performance of the detectors with the expectation of isotropic emission of particles, taking their geometry into account. Recommendations for improvements to the analysis framework and documentation are given.

Contents

1	Introduction	1
2	Background & Detector Theory	1
2.1	Coordinates	1
2.2	ALICE Run 3 Detector Array	2
2.3	The Inner Tracking System	3
2.4	The Muon Spectrometer	4
2.5	The Muon Forward Tracker	6
2.6	The Online-Offline Analysis Framework	8
2.7	Online Analysis	8
2.8	Offline Analysis	9
3	Analysing Using O2	9
3.1	AOD Structure	9
3.2	Analysis Task Structure	10
3.3	Table Features	10
3.4	Histogram Registry	11
3.5	Folder Structure	11
3.6	Compiling O2Physics	12
3.7	Running a Task with O2	13
4	Our Analysis	13
4.1	The Data	14
4.2	Initial MFT Analysis	14
4.3	Comparing pass 3 to pass 4	16
4.4	ITS Analysis	18
5	Discussion and Recommendations	22
6	Conclusion	23
A	Downloading data from the GRID	25
B	Additional Plots	26

List of Figures

2.1	Modified cylindrical coordinate system used at the LHC	2
2.2	Schematic view of the ALICE detector setup for Run 3 of the LHC	3
2.3	Schematic view of the Inner Tracking System	4
2.4	Diagram of the layout of the Muon Spectrometer	5
2.5	Schematic view of the Muon Forward Tracker	6
2.6	Pixel detector layout on the front and back plane of half-disk 4	7
2.7	Functional flow of the O2 framework	8
4.1	Gistograms of η , φ , nClusters , and z for tracks from pass 3 in the MFT	14
4.2	x - y histograms of tracks at different planes of the MFT	16
4.3	η - φ histogram for tracks from pass 3 in the MFT	17
4.4	Comparison of η distributions of tracks from pass 3 and pass 4 in the MFT	17
4.5	η - φ histogram for tracks from pass 4 in the MFT	18
4.6	Histograms of η , φ , and ITS inner barrel nClusters for tracks in the central region .	19
4.7	Histograms of η , φ , nClusters , and z for tracks in the ITS	20
4.8	η - φ histogram for tracks in the ITS	21
4.9	Histograms of cluster information for tracks in the ITS	22
B.1	Histograms of η , φ , nClusters , and z for tracks from pass 4 in the MFT	26
B.2	x - y histograms of tracks at different planes in the MFT	27

1 Introduction

The ALICE (A Large Ion Collider Experiment) detector is a detector experiment at the Large Hadron Collider (LHC) at CERN. Its primary goal is the investigation of “strongly interacting matter at extreme energy densities, where a formation of a new phase of matter, the quark-gluon plasma, is expected” [1]. It achieves this goal by studying the products of head-on collisions of heavy ions such as lead, called Pb-Pb collisions for short. It also studies proton-lead (p-Pb) and proton-proton (p-p) collisions.

Run 3 is the latest period of data capture at the LHC, with an intended centre of mass energy per collision of $\sqrt{s} = 13.6$ TeV and increased luminosity of collisions—a factor 10 increase in integrated luminosity for Pb-Pb collisions. For Run 3, ALICE is moving from a triggered read-out system to a combination of triggered and continuous read-out. In order to achieve this, many detectors and their front-end electronics were upgraded, some new detectors were added, and the analysis framework was overhauled entirely.

Of the upgrades to ALICE, the following are the subject of this report. The Inner Tracking System (ITS) was upgraded with an entirely new pixel detector technology, hoping to greatly increase the resolution when determining the primary collision vertex. The Muon Forward Tracker (MFT) is one of the new detectors added. Its primary use is to assist the Muon Spectrometer (MCH) with vertexing and tracking in the forward region of ALICE and uses the same technology as the ITS. To deal with the increased volumes of data, a new analysis framework was introduced called Online-Offline (O2).

This report aims to introduce the detectors and the data they output, trace the path of that data through the new analysis framework until it arrives at our desk, and show the details of the analysis we performed to determine if the upgrades are working as expected. The data used in this report is from two proton-proton collision runs performed in October 2021, at a centre-of-mass energy of 900 GeV. This is not an energy we expect to use for physics data analysis but is good enough for this purpose.

2 Background & Detector Theory

2.1 Coordinates

The coordinate system used at ALICE needs to be discussed in order to fully explain the scope of this report. A modified cylindrical coordinate system, shown in figure 2.1, is used as most detectors in the experiment are cylindrically symmetric about the beamline of the LHC.

We place the z -axis along the beamline with its origin at the interaction point (IP). The IP is the point at which collisions happen, right in the center of the detector. The angle around the z -axis is called the azimuthal angle, denoted by φ . Sometimes in the literature φ ranges from 0 to 2π and sometimes it ranges from $-\pi$ to π . We will try stay consistent and use the latter in this report, but we may need use the other convention at times. The angle from the z -axis in the direction of the x - y plane is called the polar angle, denoted by θ , and runs from 0 to π . We are interested in the standard 3-momentum of particles that we track in the detector, which we call $\vec{p} = (p_x, p_y, p_z)$, but we also define the transverse momentum as

$$p_T = \sqrt{p_x^2 + p_y^2}. \quad (2.1)$$

We define the rapidity, often denoted as y , as

$$y = \frac{1}{2} \ln \left(\frac{E + p_z}{E - p_z} \right) \quad (2.2)$$

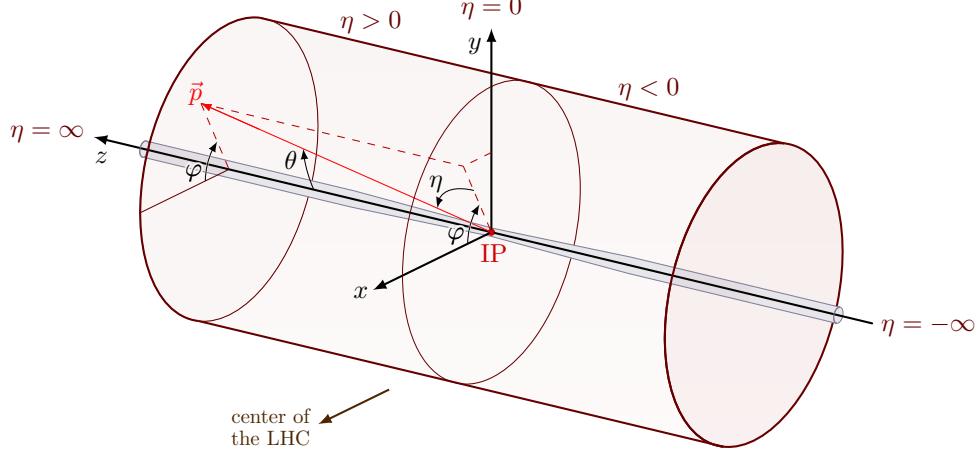


Figure 2.1: Modified cylindrical coordinate system used at the LHC [2].

where E is the total energy of the particle being considered and p_z is the momentum in the z direction [3]. This quantity is useful as differences in rapidity are Lorentz invariant for boosts along the z -axis. One issue, however, is that the energy of a particle is hard to measure without absorbing the particle entirely, so we instead use pseudorapidity, denoted as η . Rapidity and pseudorapidity are equivalent for massless particles and near equivalent for particles with total 3-momentum magnitude p much greater than their mass m . Pseudorapidity is much easier to measure as it is defined in [3] as

$$\eta = -\ln \tan \frac{\theta}{2}. \quad (2.3)$$

From figure 2.1 we see that for z positive, η is also positive, and similarly for z negative. Confusingly, we define the “forward region” of the ALICE detector as the region for which z , and thus η , are negative. The “central region” is roughly defined as $|\eta| < 1$. The four coordinates that we use most often are z , φ , p_T , and η .

2.2 ALICE Run 3 Detector Array

In 2018, the LHC shut down for what was called Long Shutdown 2 (LS2). During this time, the ALICE experiment was being prepared for Run 3, where it will take data at higher energies and much higher luminosities than before, from 2022 until 2025 [4]. Figure 2.2 shows the detector configuration for Run 3. The intent of these upgrades was in large part to prepare ALICE for a higher luminosity of collisions in both Pb-Pb and p-p cases.

Part of the upgrades for Run 3, the details of which can be found in [4], were a whole new Inner Tracking System and a brand new detector called the Muon Forward Tracker. These detectors are both silicon-based and their primary purpose is tracking particles and determining the collision vertex, which is the best estimation of where the collision that resulted in these particles happened.

The read-out electronics for many detectors were upgraded to allow for continuous read-out where necessary. The MCH also had its read-out and front-end electronics upgraded but will still work on a triggered read-out system.

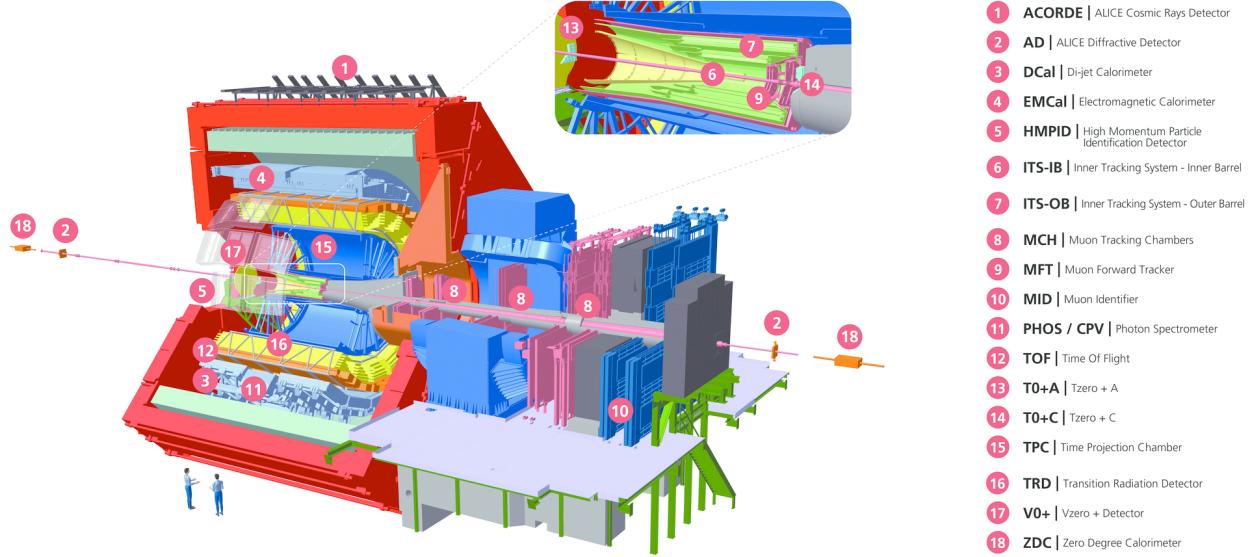


Figure 2.2: Schematic view of the ALICE detector setup for Run 3 of the LHC [5]. Note here that the MCH is shown separate from the MID, which act as the triggering mechanism for the MCH. For the purposes of this report, the MID will be considered part of the MCH. The ITS (6, 7), MCH (8), and MFT (9) are the focus of this report. The large red casing is a solenoid magnet whose field lines point along the beam axis and whose direction can be flipped or turned off entirely.

2.3 The Inner Tracking System

The Inner Tracking System (ITS) sits in the main barrel of ALICE, as seen in figure 2.2, and covers the range $|\eta| < 1.22$ [6]. For Run 3 it has been upgraded significantly by replacing the old detector with a new layout and new pixel detector technology, leading to an improvement in track position resolution at the primary vertex of a factor of 3 or greater [6]. The ITS's main purpose is to track the particles resulting from the collisions and determine the position of the primary vertex of collisions. It also serves to “reconstruct secondary vertices, track and identify particles with low momentum, and improve the momentum and angle resolution for particles reconstructed by the Time Projection Chamber (TPC)” [7].

The new ITS consists of 7 layers of pixel detectors; 3 in the “Inner Barrel” and 4 in the “Outer Barrel”. The innermost layer sits at a radius of only 22.4 mm from the IP thanks to a reduction in beam pipe radius for Run 3 and the outermost layer sits at a radius of 391.8 mm from the IP. Figure 2.3 shows the layout more clearly.

The pixel detectors used are $0.18\text{ }\mu\text{m}$ CMOS chips from TowerJazz. When a charged particle passes through the silicon in the active volume, it liberates the charge carriers in the material, which then collect on electrodes connected to the silicon, telling the detector that a particle has been detected. The fine segmentation of the detectors also allows the detector to determine the point at which the particle hit the detector, up to a resolution of $4\text{ }\mu\text{m}$ in both the $r\varphi$ and z directions [6]. The amount of charge deposited on the detector is dependent on the particle species and momentum (Bethe-Bloch).

The ITS creates particle tracks in two stages. Firstly it looks at tracks found in the TPC and uses them as seeds to create tracks from the clusters of hits in the ITS with a Kalman filter [8], thus extending the TPC tracks and giving them much better timing information. This method works for particles with $p_T \gtrsim 100\text{ MeV}/c$ but below that, the TPC acceptance drops off considerably so

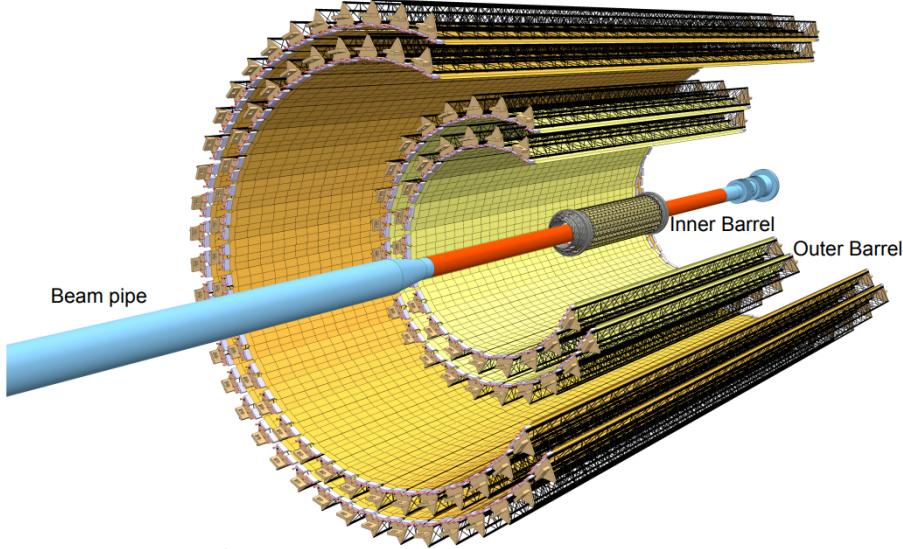


Figure 2.3: Schematic view of the Inner Tracking System [6]. Note the thinner beam pipe and extremely close Inner Barrel, as well as the structure of pixel chips running along the beam line.

stand-alone ITS tracking takes over, working on those clusters that were not used in the previous method. The Cellular Automaton method, described in more detail in section 2.5, is used for this stand-alone track finding. There doesn't appear to be a minimum number of layers required for a track to be accepted in the stand-alone regime but the TPC regime requires 5 out of 7 layers contributing.

Two main methods of read-out were considered for the ITS in the new continuous read-out scheme. Firstly, a rolling shutter which continually loops through the rows of pixels and reads out the charge deposited on that pixel in the time since the last read-out was considered. The time between readings, known as the integration time, for the first method is around $30\text{ }\mu\text{s}$. The rolling shutter scheme lends itself to needing a small number of transistors within each pixel. The second scheme is known as ALPIDE, where each pixel has a comparator that signals when the pixel has an analogue signal greater than the comparator's threshold. The signalled pixels then get read out asynchronously, according to their priority in the chain. This scheme has an effective integration time of around $4\text{ }\mu\text{s}$ but has a larger material budget. At time of writing it is not clear to us which read-out architecture is used in the ITS.

2.4 The Muon Spectrometer

The MCH sits in the forward region of ALICE, as seen in figure 2.2, and covers $-4 < \eta < -2.5$. It is designed to study heavy quark resonances through their single- and di-muon decay channels. As is shown in figure 2.4, it is composed of a hadronic absorber, 5 tracking chambers, a dipole magnet, another absorber, and finally the 2 trigger chambers. The MFT is often also considered part of the MCH but is not shown in figure 2.4. This section is adapted from the MCH Technical Design Report [9].

Muons don't tend to interact with matter much, especially when compared to hadrons and electrons. This makes studying muons easier, to some degree, since most other particles can be filtered out by making them pass through a large chunk of matter. As is shown in figure 2.4, that

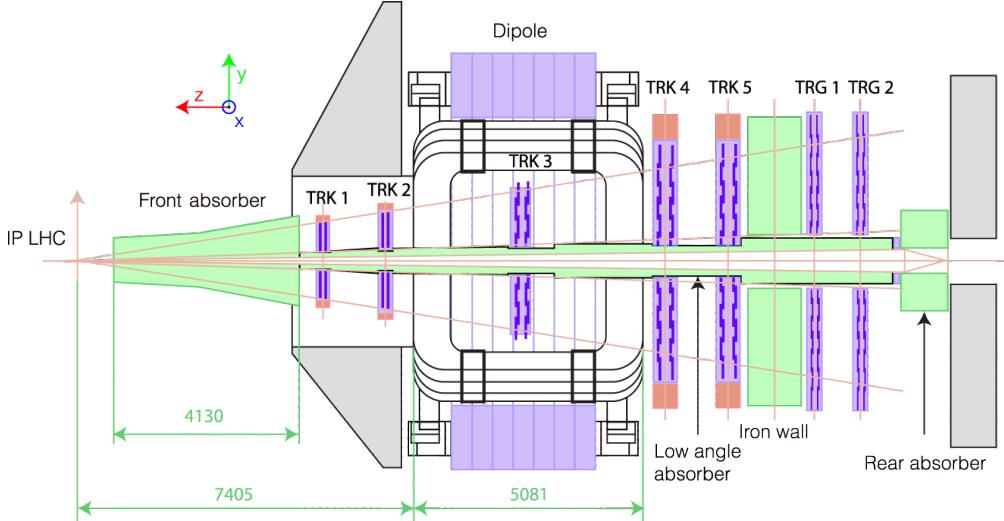


Figure 2.4: Diagram of the layout of the Muon Spectrometer [10]. Muons pass through the absorber, are deflected by the dipole magnet, and hit the trigger chambers at the back. Importantly, all detector material sits behind the hadronic absorber, meaning the amount of data that can be used for tracking and vertexing is much lower than, say, the ITS.

is exactly what is done in the MCH. In front of any detector material (ignoring the MFT for now) sits the hadronic absorber, made primarily of carbon and concrete. This is intended to filter out all non-muon particles (mainly hadrons and photons) while not reducing the muon energy much so that they can still be studied.

After the absorber are 5 sets of 2 cathode pad tracking detectors, situated around a large warm dipole magnet. Particles that make it through the absorber get picked up by the first two sets of detectors, then pass through the magnetic field and are deflected according to their charge, mass, and momentum. The third set tracks the particles during deflection and then the last two detect them after deflection. This set-up is particularly useful for studying di-muon events as the muons produced would be a muon-antimuon pair, which would deflect in opposite directions in the dipole magnet. This would leave a characteristic track signature that can be studied.

After the last tracking detector, particles pass through another absorber, which serves to further filter out muons from background, as well as filter out background muons. The muons produced in heavy quark resonances have considerably higher p_T than those produced by background processes, so the job of the trigger system is to only trigger on muons with high enough p_T to be interesting (this is defined per process) and the second absorber helps reduce the number of background muons incident on the trigger system. The trigger system is made of 2 stations of 2 Resistive Plate Chamber (RPC) detectors. Comparing the measurement of the same particle in the two stations, the p_T can be determined. The decision to keep or reject an event takes about 300 ns.

In Run 1 and Run 2, the MCH performed all its own tracking and vertexing on the particles it studied. Particularly for vertexing, where the collision position is estimated, this was not optimal as most of the particles produced in the interactions didn't make it through to sensitive material. With the increased energy and luminosity of Run 3 a better system was needed to perform these tasks, so the MFT was added in front of the first absorber to take over the job.

2.5 The Muon Forward Tracker

The MFT is a brand new detector added to ALICE for Run 3 to assist the MCH with tracking and vertexing. It covers the range $-3.6 < \eta < -2.45$ and was designed in conjunction with the ITS, using precisely the same CMOS pixel detectors. Due to the MFT being placed in front of the absorber, it detects a lot more particles than make it through to the MCH, allowing it to be much better at finding the primary vertex of collisions. Figure 2.5 shows the design of the MFT. The rest of this section is adapted from the MFT Technical Design Report [11].

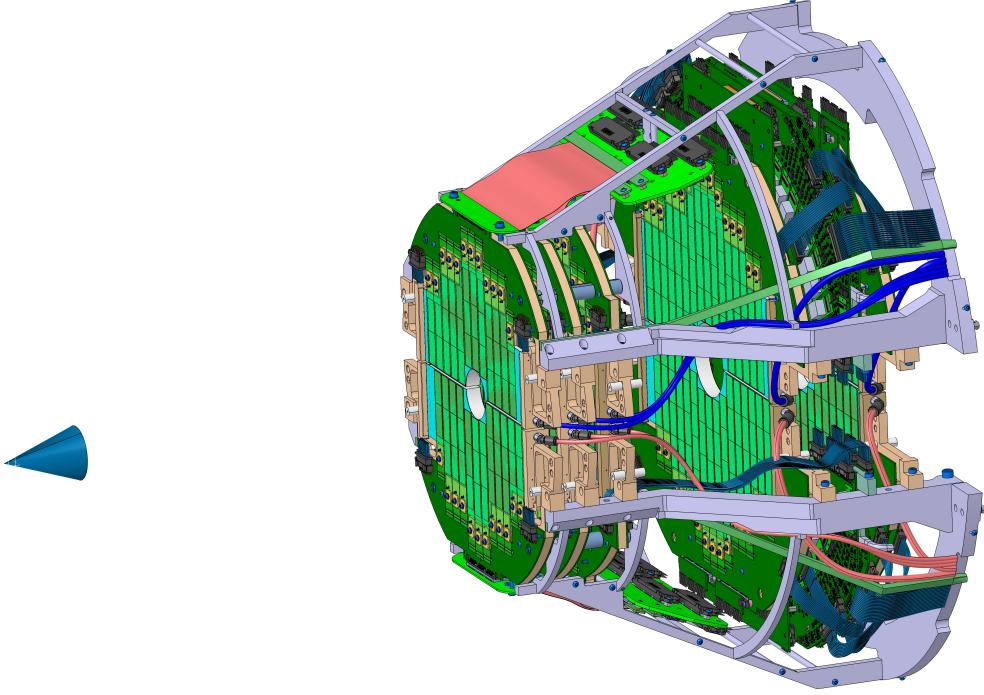


Figure 2.5: Schematic view of the Muon Forward Tracker [12]. The small cone on the left shows the IP. Note that the 5 disks each have a front and back plane of pixel detectors, totalling 10 z -positions for the MFT to “see” particles at. The slight gap between top and bottom half-disks can also be seen.

The MFT is made of two identical half-cones sandwiching the beam pipe from above and below, each with 5 half-disks positioned at different distances from the IP along the z -axis. Each half-disk has a front and back detection plane totalling 10 detection planes. As the disks get further from the IP their radius increases in order to cover the same η range, aside from the second disk, which is identical to the first. The disks sit at z -positions -46.0, -49.3, -53.1, -68.7, and -76.8 cm respectively and each disk is 1.4 cm thick, leading to detector planes at ± 0.7 cm from each of those positions.

The half-disks are made from ladders of 2 to 5 rectangular pixel detector elements. Figure 2.6 shows an example of the layout of the front and back planes of detector elements in both x - y and η - φ .

The MFT performs stand-alone track reconstruction in two steps. “Track finding” filters through clusters (groups of hits in a certain layer) to group into track candidates and then “track fitting” fits tracks to those clusters and determines kinematics and covariance matrices. A Kalman filter [8] is used for the track fitting but while it could also be used for the track finding, a more advanced algorithm is used to increase efficiency and reduce computation time.

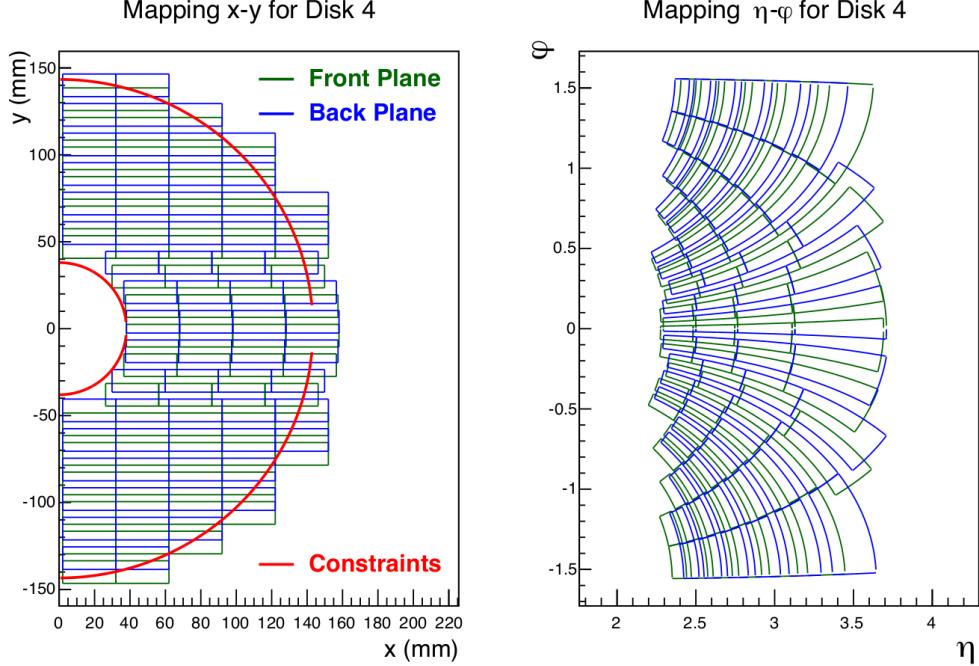


Figure 2.6: The layout of pixel detector elements on the front and back plane of one of the half-disks of disk 4 in the MFT. Note that the front and back plane have layouts offset to each other by half the width of the pixel elements. This figure is taken from [11, fig. 6.1] but we would like to point out that the axes are labelled incorrectly in both cases; on the left, x and y should be swapped as the ladders run vertically and on the right φ should run from $-\pi$ to 0 while η should be negative.

There are two algorithms used in conjunction to group clusters into track candidates. The first, called the Linear Track Finder (LTF) Algorithm, assumes that, in the η region that the MFT occupies, the solenoid magnet surrounding the central barrel has a reduced bending effect on tracks, leading to effectively straight tracks. A “seeding line” is determined from clusters in the first and last disk, often with the help of the position of the primary vertex, and the algorithm simply minimises the distance of clusters to that line. There is a single parameter for this algorithm which is the radius around a cluster to allow a seeding line to be considered. While the LTF method is very computationally inexpensive, its straight line assumption breaks down at low momenta, so the Cellular Automaton (CA) algorithm is implemented to pick up the pieces.

The CA algorithm works by pairing clusters in neighbouring disks into “cells” and then iterating through cells and determining, based on a number of parameters such as maximum deviation angle and maximum θ angle of a cell according to the collision vertex, whether neighbouring cells are “compatible” and grouping those cells together as candidates for a track. This algorithm depends on having a determined position of the collision vertex to both create cells and to match cells together as in both cases they need to point towards the vertex within some pre-defined limit. This vertex position can be supplied by external means, such as the ITS, or it can be estimated using cells between the first two disks. The algorithm also has a default minimum track length in terms of number of disks involved in the process, which is set by default at 4 out of 5 disks.

Due to the number of cluster combinations that the CA algorithm needs to consider compared to LTF, it ends up being considerably more computationally expensive to run. Since both methods have their advantages, particularly with CA picking up the slack in the low momentum (and

specifically low p_T) regions, both methods are implemented. LTF goes first, finding as many high p_T tracks as it can, and then CA runs on the clusters that the LTF did not find tracks for. This reduces the overall computation time and takes advantage of both algorithms as much as possible.

2.6 The Online-Offline Analysis Framework

With the increased interaction rate intended for Run 3, a new system for real-time processing, as well as offline analysis, needed to be constructed [4]. The Online-Offline (O2) framework was developed for this purpose. This section is mostly adapted from the O2 Upgrade TDR [13] as well as the O2 documentation at <https://aliceo2group.github.io/analysis-framework/>.

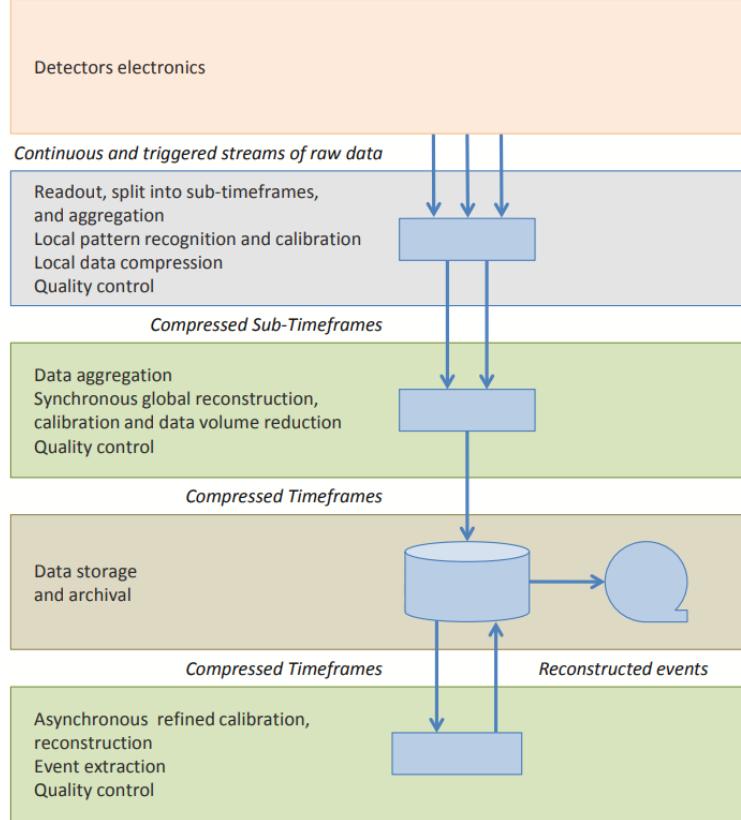


Figure 2.7: Functional flow of the O2 framework [13]. Detectors output their signal continuously in the new configuration and this signal gets split into chunks called timeframes. These timeframes get processed a number of times, reducing the volume of data each time by only extracting the quantities that will be used for analysis. Many choices have to be made at each step to ensure that useful data makes it out the other side and so the details of this process are always in flux.

2.7 Online Analysis

The “Online” portion of the framework is the part that does all the synchronous processing, i.e. as the raw data is continuously read out from the detectors. The data is first stored in “sub-time frames” (STFs), which get processed to generate clusters in order to reduce the volume of data and then get built into “time frames” (TFs) of 10 ms chunks of data. From there the data gets its first pass of detector reconstruction, e.g. track finding, and then gets compressed into “compressed

time frames” (CTFs). The CTFs are stored on disk and that is where the Online portion ends. Note that at each step, QC and calibration data is extracted and stored for later use.

2.8 Offline Analysis

With the CTFs stored to disk, they can then be processed asynchronously into Event Summary Data (ESD) and then Analysis Object Data (AOD) files. ESDs are not used for analysis and are stored for a while before being deleted. AODs (real and simulated) and CTFs are the only persistent data type.

The process of turning CTFs into AODs is called “reconstruction” and is often performed multiple times on the same data. Iterations of reconstruction are called reconstruction passes and these passes will usually be done many times if the previous passes missed something in their reconstruction, or perhaps if a certain table was not populated in the reconstruction but someone needs that table for their analysis. As a result of this, two different passes on the same data can end up with very different-looking results.

We distinguish between data taken in Run 3 and data taken in Runs 1 and 2 by calling Run 3 AODs “AO2D” files, and Run 1 and 2 just “AOD” files. The data is stored on the `alimonitor` system, requiring a browser certificate to access, which is obtained by joining the ALICE collaboration. Access to all data and most analysis tools used in this report is restricted behind this wall. AODs can then be analysed using an “Analysis Task”, which is written in C++ and ROOT.

The focus of the upgraded analysis framework was to reduce disk space usage when processing and analysing, as well as making sure all analysis takes advantage of all processing power available to it at all times. These themes are present in every step.

3 Analysing Using O2

One of the main goals of O2 is to take away as much responsibility from the user as possible when it comes to managing memory, writing efficient code, and disk space usage. To this end, writing an analysis task in O2 is structured quite strictly. This section will outline the information needed when trying to analyse reconstructed data to produce, in our case, histograms of kinematic variables.

It must be noted that learning how to do this, and how to deal with the idiosyncrasies of the O2 software, is what took up the majority of the time spent on this project. O2 is written to do one thing very, very well, but unfortunately that comes with the side-effect of it being extremely picky about the conditions in which the software will actually work. Lastly, a distinction needs to be made between O2 and O2Physics. O2 generally refers to the entire analysis framework, including both the online and offline parts. This encompasses O2Physics, which is used only for offline analysis. Despite this, in terms of the structure of the software, O2 and O2Physics are separate things but O2Physics sometimes depends on O2. For this reason it can become ambiguous when discussing O2 and O2Physics but we will try our best to keep them distinct.

3.1 AOD Structure

The data that we use in our analysis comes in the form of Analysis Object Data. These get produced asynchronously and come in the form of ROOT files containing “Dataframes”. These are organised in a tree structure where each tree contains a number of tables, for example a table containing collisions and a table containing tracks. The tables have columns which hold variables corresponding to each entry, or row, in the table. There are 4 types of column:

- Static columns are saved to disk during the reconstruction process and are available at any time. The z -position of a collision vertex, for example, is a static variable.
- Dynamic columns are defined as functions with inputs from static variables (or variables input by the user). They will produce a result on demand in order to save memory and disk space. Momentum in the x -direction is a dynamic variable.
- Index columns point from one table to another, such as from a track to its associated collision. They can also refer to their own table, such as for a simulated particle to refer to its mother or daughter particle after decay.
- Expression columns are similar to dynamic columns in that they get calculated on demand, but they get evaluated on all entries in a table upon requesting it and produce a column that can then be accessed as if it were a static column. Importantly, dynamic columns cannot be used as inputs to expression columns.

3.2 Analysis Task Structure

Analysis tasks are written in C++ and need to be structured in a specific way so that O2Physics can use them properly. Each task is written as a `struct` object which is then called at the end of the task. Below is an outline of what is needed for a task.

```

1 #include "Framework/runDataProcessing.h"
2 #include "Framework/AnalysisTask.h"
3
4 using namespace o2;
5
6 struct MyTask {
7   // Define things here, such as histogram registries, filters for data, or new tables
8
9   void init(framework::InitContext&)
10  // Here we initialise histograms and other things used in the analysis
11  {};
12  // The arguments of the process function are where we subscribe to specific tables in the AOD
13  // → that we're analysing.
14  void process(aod::Collision const& collision, aod::Tracks const& tracks)
15  // Here we can do any processing that we need, calculating things etc, and then fill the
16  // histograms we defined earlier
17  {};
18
19 // This is what O2 looks at to run the task
20 WorkflowSpec defineDataProcessing(ConfigContext const& cfgc)
21 {
22   return WorkflowSpec{
23     adaptAnalysisTask<MyTask>(cfgc),
24   };
}

```

3.3 Table Features

All variables associated with a track, for example, could be included in a single table. However, since we often only need a few of the variables, the tables are split up into sections that contain variables often used together. If needed, these tables can be joined together when doing analysis

using `o2::soa::Join<Table1, Table2>`. Importantly, only tables which correspond row-to-row and have the same number of rows can be joined in this way.

We might also want to only access entries in a table that pass some filter, say having $|\eta| < 0.8$. We can do this by putting `Filter f = nabs(aod::track::eta) < 0.8;` somewhere before the `process` function and then instead of subscribing to `aod::Tracks` as we have done, we can put `soa::Filtered<aod::Tracks> const& filteredTracks` in the `process` function call. Multiple filters can be defined and only compatible tables will be filtered. Once a table has been filtered, the unfiltered version cannot be accessed.

Similar to filters are partitions, which we define like `Partition<aod::Tracks> leftEta = aod::track::eta < 0;`. This will create a separate table that we can subscribe to which will only contain tracks with the specified η value. This is different to a filter as the original table can also be accessed at any time.

Some tables have pre-defined iterators which can be used to more efficiently loop through all entries in the table. `Collisions`, for example, has an iterator called `Collision` that we subscribed to in the code example above. When a table has an index column relating to `Collisions`, as `Tracks` is, O2 will automatically group the entries in `Tracks` by associated collision.

3.4 Histogram Registry

Histograms are the eventual desired output of an analysis task and O2 has a class designed specifically to make creating and filling histograms easier. Before the `init` function we can define our histogram registry and then in the `init` function we can define the histograms we want and add them to the registry. We can then fill the histograms in the `process` function.

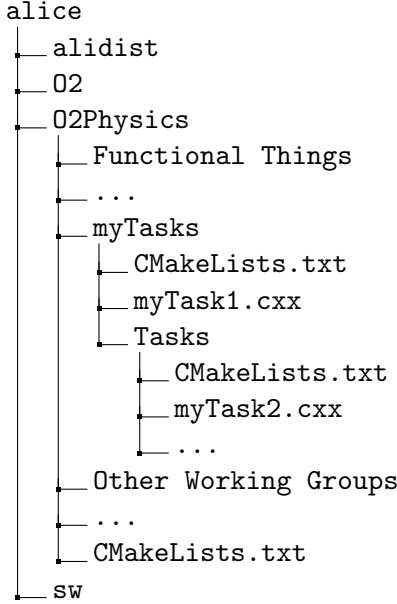
```

1 HistogramRegistry myRegistry{ // This name and the name in the next line need to be the same
2   "myRegistry",
3   {},    // Histograms could be defined here but we will do it in init()
4   OutputObjHandlingPolicy::AnalysisObject, // Tells the task which format to output in
5   true,   // Sorts the output histograms alphabetically
6   false   // Won't create a subdirectory for this registry. Set to true if more than one registry
→   is being used
7 };
8
9 void init(o2::framework::InitContext&){
10   AxisSpec etaAxis = {nBins, binMin, binMax, title} // title is what gets printed on the axis
11
12   myRegistry.add("myHist", "myHist", kTH1F, {etaAxis}) // Initialises the histogram. First
→   argument is the name of the histogram, to use internally, and second gets printed as the title
13 }
14
15 void process(aod::Tracks const& tracks) {
16   for (auto& track : tracks) {
17     registry.fill(HIST("myHist"), track.eta()); // Fills the histogram with eta from all tracks
18   }
19 }
```

3.5 Folder Structure

With the task written, it then needs to be compiled and added to O2Physics so that it can be run. O2Physics has a number of analysis tasks written by people at ALICE which get compiled automatically. These are sorted into physics working groups such as Heavy Flavour (PWGHF) and

Jets (PWGJE). If we want to add our own task to O2Physics, we need to create our own folder with the same structure as the working groups. Below shows the structure of the file system.



Here `alidist` is the git repository that handles the versioning of O2 and O2Physics. O2 is what handles the backend of the analysis framework, compiling the tasks written in O2Physics. The `CMakeLists.txt` files are needed at every level of the O2Physics structure to tell O2 what to compile and which commands to use to refer to things.

See below an example of what the `CMakeLists.txt` file in the `myTasks` folder would look like.

```

1 add_subdirectory(Tasks)           # Ensures O2 can see the Tasks folder
2 o2physics_add_dpl_workflow(my-task1) # The command assigned to the task. Note only lowercase
→ letters, hyphens, and numbers are allowed
3     SOURCES myTask1.cxx          # The source file for the task
4     PUBLIC_LINK_LIBRARIES O2::Framework
5     COMPONENT_NAME Analysis)

```

3.6 Compiling O2Physics

O2 and O2Physics are built using `aliBuild` [14]. They prefer to be built on UNIX systems and require at least 8 GB of RAM, preferably more. We used both O2 and O2Physics as we wanted to create and run analysis tasks.

Once they are built (those four words are doing some *very* heavy lifting) we can enter the O2Physics environment with `alienv enter O2Physics` and this will place us in a new terminal shell. The magic of the software is that it compiles all the analysis tasks in O2Physics, as well as tools for simulation and the like, such that everything can be done by running commands in that shell. Before running our own tasks, however, we need to tell O2 to build our tasks into O2Physics. To do this we use `ninja`.

If we have our tasks written and files structured as shown in section 3.2, we can enter an O2Physics environment and load `ninja` alongside using `alienv enter O2Physics ninja/latest`. In the shell we can then navigate to the build of O2Physics, which should be in `alice/sw/BUILD/O2Physics-latest/O2Physics` and run `ninja install myTasks/all` which will rebuild only those parts of O2.

Ninja is the quick and dirty way of doing things so every now and then it's a good idea to rebuild O2 and O2Physics entirely, pulling the latest release. We can do this while also making sure our own tasks don't get overwritten using the following steps. We first need to make sure that git knows our files are there with `git add path/to/files` and then committing with `git commit -m "commit message"`. With that done, we can systematically update `alidist`, `O2`, and `O2Physics` by navigating to each and running `git pull --rebase`. Finally we can rebuild by navigating to `alice` and running `aliBuild build O2Physics --defaults o2 --debug`. This will take a few hours to complete (if it's even successful) and then will be able to be used again.

3.7 Running a Task with O2

Once we have our tasks created and built in O2Physics, we can then run them. For our purposes, the only commands we need to know are how to run a task and the options that come along with that. All analysis tasks in O2Physics get assigned a unique command that can be used to run that task. They all begin with `o2-analysis...` followed by the name assigned to it in the relevant `CMakeLists.txt` file as shown in section 3.5. In the case of that task, we would run it with `o2-analysis-my-task1`.

Most analysis tasks are run on `AOD.root` or `A02D.root` files so in order to tell the task which file to use, we use the flag `--aod-file A02D.root`. We could also supply a list of files in a text file and use `--aod-file @A02D_list.txt` where `A02D_list.txt` contains the path to the files we want to run on, one on each line.

Often we want to run multiple tasks in succession on the same data, feeding the output of one into another. To do this, we simply use the pipe symbol `|` between the tasks: `o2-analysis-track selection | o2-analysis-my-task1 --aod-file A02D.root`. Here the ordering of the tasks doesn't matter as the input and output format of a task is known before it runs, so O2 does some quick thinking to arrange the workflow such that the tasks get fed the correct format of data.

The `o2-analysis-trackselection` task above is an example of a helper task. These are specifically designed to use the available data to produce tables that are needed in analysis tasks. This is done once again in the interest of saving disk space as calculating the values needed is much less resource-heavy than storing them for all time. A list of helper tasks and the tables they produce can be found at <https://aliceo2group.github.io/analysis-framework/docs/datamodel/helperTaskTables.html>. Note that when running a workflow with multiple tasks, if a configuration file is needed, or some other option, the flag needs to be provided before the pipe symbol for each task that gets called.

For the most part, the output of an analysis task is either a `AnalysisResults.root` file or a `QAResults.root` file, with the former being the most common. This output type is chosen when defining the histogram registry. If all goes well (which is a rare occurrence) the ROOT file produced can be opened with TBrowser and some lovely histograms should pop up. These can be saved as is or output using ROOT macros.

4 Our Analysis

The intention of this report is to investigate preliminary data coming from the ITS and MFT in Run 3. The order of the discussion in this section will follow the order that we tackled things. This is done to show both the progression of our knowledge as well as to try to clarify some of the explanations in previous sections as the only way to truly understand some things is through example.

4.1 The Data

The data analysed in this report was taken in October 2021, where protons were collided at a centre of mass energy of 900 GeV. This is not an energy that we expect to use for physics research but it allows us to look at how the detectors are performing with more lightweight data, simply because there will be fewer particles created in the collisions and thus less data to work with. We are using runs 505548 and 505645. In this case, a run specifies a period of data taking for which all global settings remain the same. All further plots, unless specified otherwise, will include both runs. The data was downloaded from the ALICE GRID using the script in appendix A.

4.2 Initial MFT Analysis

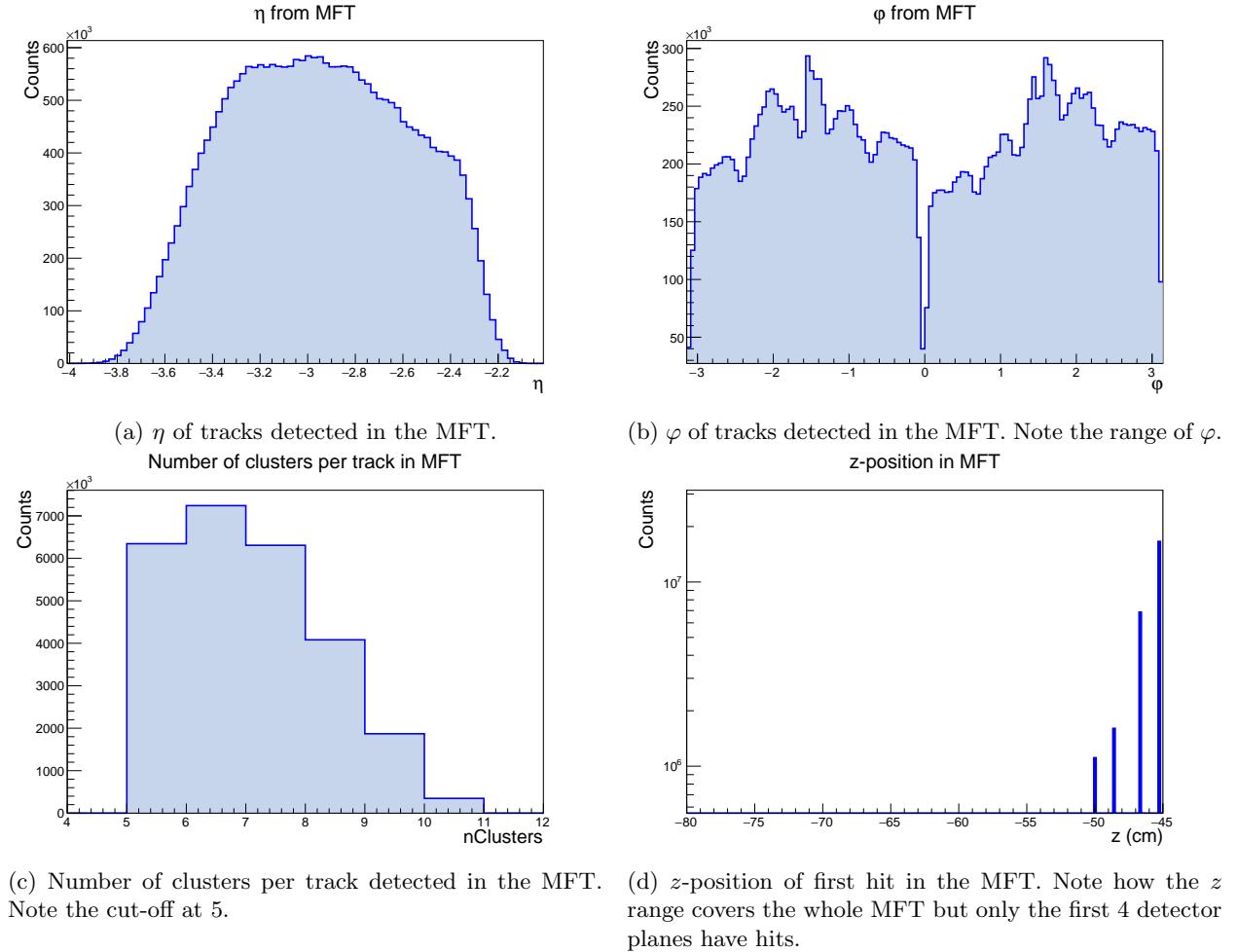


Figure 4.1: 1-D histograms for some kinematic variables as well as the number of clusters per track in the MFT. Data from reconstruction pass 3.

In the AOD data model there is a table called `aod::MFTTracks` which contains the tracks detected in the MFT. When we began this analysis, only two reconstruction passes had been run on the data and while the MFT was switched on for the runs, the `aod::MFTTracks` table had not been populated. We thus had to wait for pass 3 and we spent that time getting familiar with the analysis framework as described in section 3. With pass 3 available, the plots in figure 4.1 were

able to be created. The variables plotted are all available as static or dynamic columns in the AOD structure so no complicated analysis was needed to obtain them.

Starting with η , we see the distribution is not as smooth in the middle as we might expect from a detector with continuous sensitive material in those regions. We will investigate this more in section 4.3. The φ plot shows the structure of the MFT quite well. The valley at the centre shows the gap between the top and bottom half-disks and the spikes come from the fact that the sensitive area of each disk is not perfectly circular, so there will be directions that can detect more tracks than others.

The number of clusters per track is interesting. As was mentioned in section 2.5, the MFT requires a track to be detected in 4 out of the 5 disks, by default, in order to be considered a track. The translation of that statement into number of clusters per track is a bit unclear but we might expect that the minimum number of clusters per track should be 4 as each disk has 2 planes but the track only needs to have a cluster in one of them to count towards the 4 out of 5. However, what we see in figure 4.1c is a minimum of 5 clusters per track. This might be down to a choice made in the reconstruction process to instead require that all 5 disks detect the track before it is accepted. It could also be due to a physical limitation where it's simply not possible for 4 disks to detect a track and not see clusters in 5 disks. At the point of writing, we have not been able to determine the reason for this as the details of reconstruction are near impossible to find.

Interpreting figure 4.1d was its own adventure. The documentation for the AOD data model is unclear about what gets calculated for the z column in the `MFTtracks` table, but we believe that it represents the z -position of the first hit (or cluster) of a given track in the MFT. We see that after the fourth plane, i.e. the second disk, there is no data. This supports the fact that 4 out of 5 disks are required to accept a track as having a first hit in the third disk will obviously never result in a track hitting 4 disks. This information seems to be contradicting the information from figure 4.1c but we regrettably have no resolution to the situation.

Next, we looked at the x and y positions of the hits in the first 4 layers that we saw in figure 4.1d. This is shown in figure 4.2 and the structure of the ladders of pixel detectors can clearly be seen. As with the z plot, there is no further data for the other layers in the MFT. It would be useful to be able to see the positions of all of the hits for a track, but we are restricted to what is added to the AOD and at this point, this is all we have to work with.

The blank regions in the x - y plots are likely areas where the data has been masked in the reconstruction process. This is most likely to be because the detector had high noise levels in those areas, which would blow out the rest of the data. Looking at the first two plots, we can see how the front and back planes of the first disk are offset from each other, as we saw in figure 2.6. We can also see an increased number of hits closer to the centre of the disks. This can be explained by the fact that, assuming particles are emitted isotropically from the IP, there will be a drop-off in density of particles as the distance from the IP increases and since the disks are flat, the larger the radius on the disk, the further its outer points are from the IP. Thus the density of hits should decrease as the disk radius increases.

The last interesting plot from reconstruction pass 3 is a 2-D histogram of η and φ . Figure 4.3 shows this and we can immediately see the resemblance to the layout in figure 2.6. As with figure 4.1b, we can see the gap in between the half-disks.

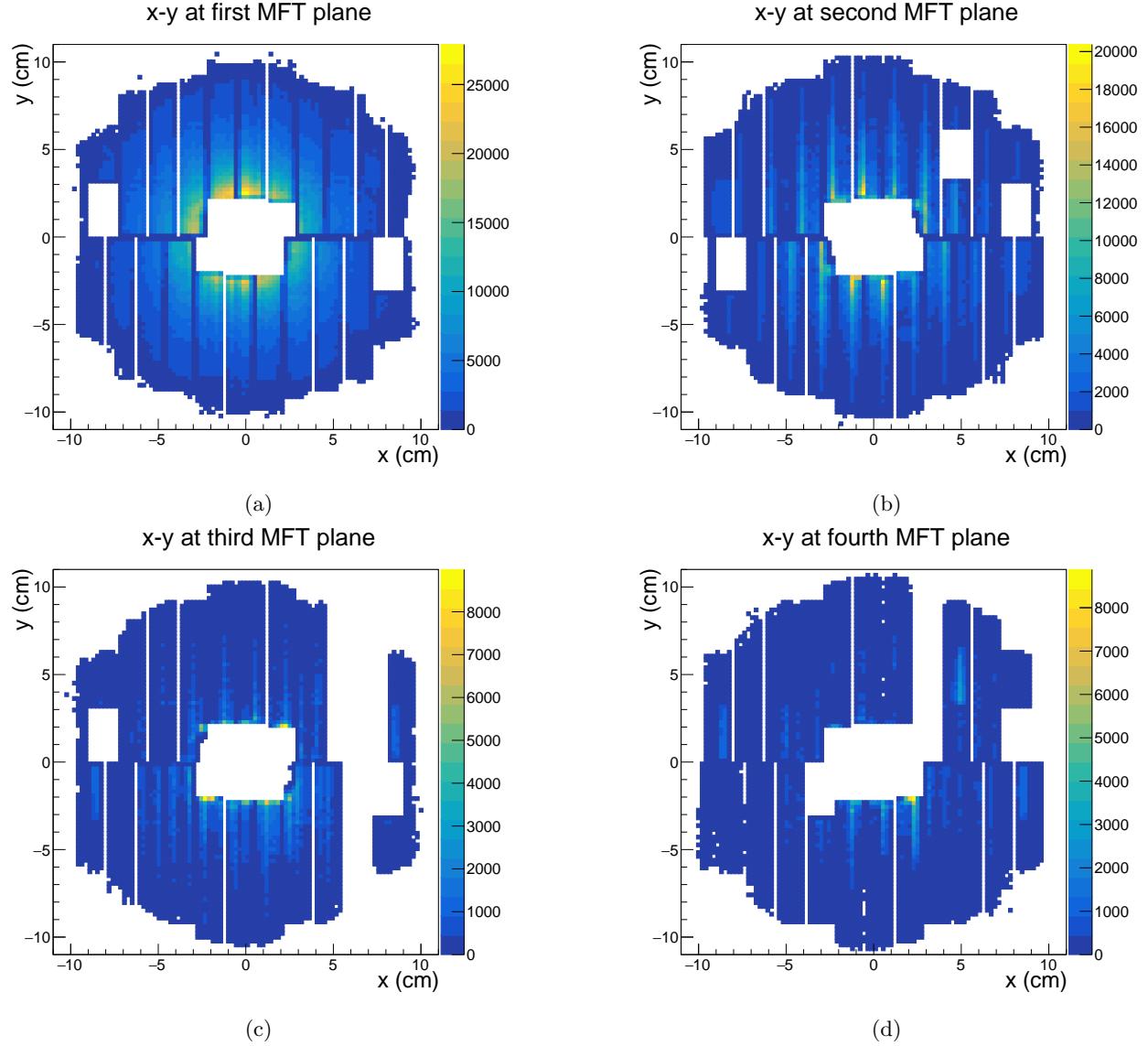


Figure 4.2: x and y positions of hits in the first 4 layers of the MFT. This is related to figure 4.1d as these are the first hits for a given track, so looking at the later layers yields no data. The sections with no data have likely been masked out to remove regions of abnormally high noise. Important to note how at the centre of the disk we see a higher hit density than further out.

4.3 Comparing pass 3 to pass 4

In figure 4.1a we can see that the distribution looks slightly jagged from about -2.4 to -3.4, especially when compared to the section between -3.4 and -4. There isn't any reason to expect a non-smooth distribution there, and that viewpoint is supported by looking at the distribution after reconstruction pass 4. Some issues were identified with pass 3 so pass 4 was performed and we have plotted the two η distributions in figure 4.4 for comparison.

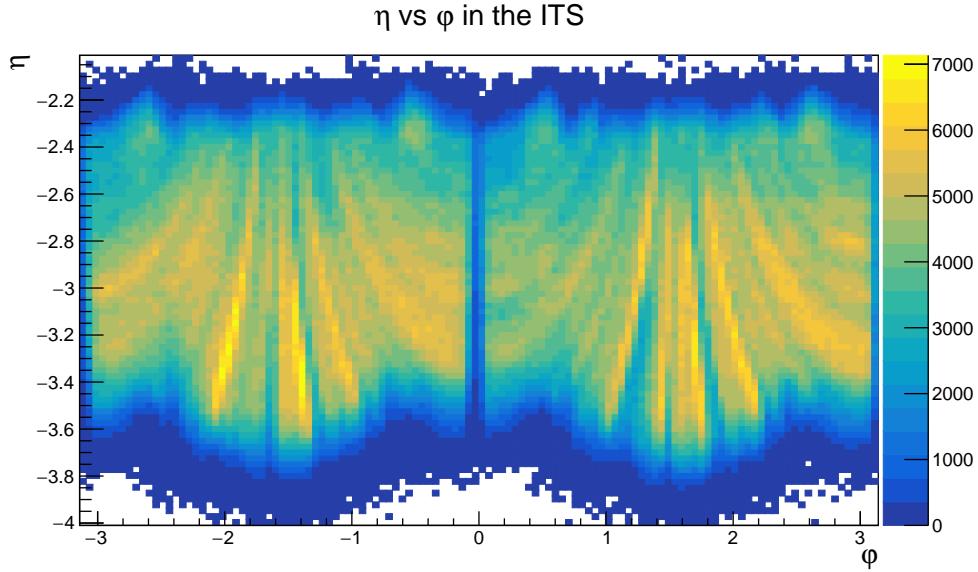


Figure 4.3: Histogram of η and φ for tracks in the MFT from reconstruction pass3. The half-disk pixel chip structure shown in figure 2.6 can clearly be seen.

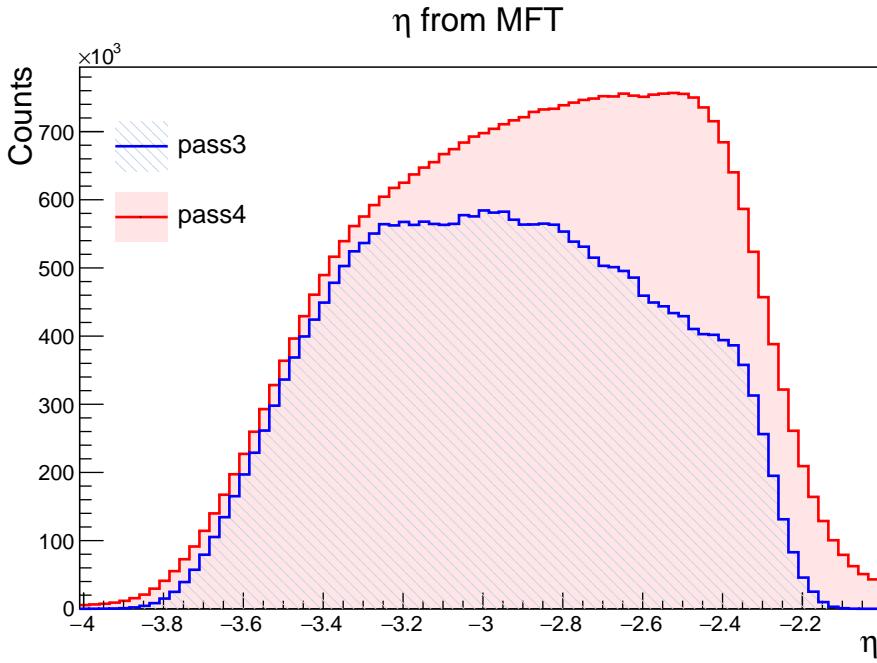


Figure 4.4: Comparison of the distribution of η per track for reconstruction pass 3 and pass 4. We notice that there is an overall increase in number of tracks, as well as a large number of tracks added in one region, leading to an overall more smooth distribution in pass 4. Note also how there seems to be a higher density of tracks in the small η region before it drops off outside the range of the MFT.

What is evident from figure 4.4 is that pass 3 seems to have missed a large chunk of tracks in one specific η region compared to pass 4. The reason for this is unclear as the details of

reconstruction are very hard to find but we suspect that it is due to pass 4 considering MCH tracks and connecting them to MFT tracks. These new tracks, and resultant η distribution, seem to contradict the conclusion drawn from the x - y plots in figure 4.2, where we saw a higher density of tracks closer to the centre of the disks, i.e. with a larger rapidity. This tension is resolvable through a discussion of the definition of η . We define η in equation (2.3) and it is very clearly a nonlinear function of θ . This means that two slices of η that have the same size will not necessarily cover the same size range of θ and since we expect tracks uniformly distributed in θ , we shouldn't necessarily expect a uniform distribution in η . The smoothness, however, is expected and pass 4 seems to have a very smooth distribution.

Lastly in comparing pass 3 to pass 4, we can have a look at the η - φ histogram, noting the more uniform distribution overall while still showing the structure we expect from figure 2.6. The uniformity in the histograms for pass 4 is an indicator that the detector is working properly as we expect isotropic emission of particles from the IP. More histograms from pass 4 can be found in appendix B.

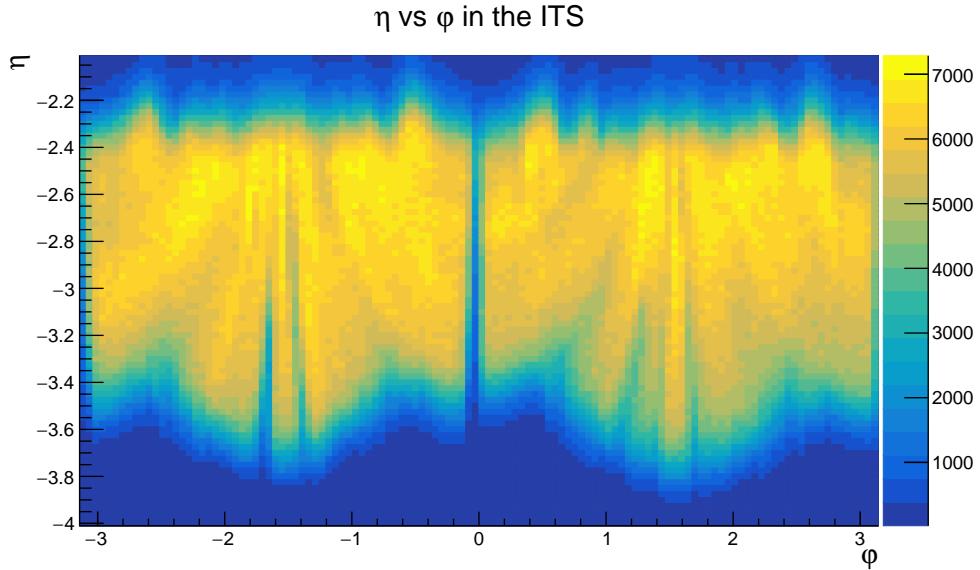


Figure 4.5: Histogram of η and φ for tracks in the MFT from reconstruction pass 4. The overall increase in tracks can clearly be seen, as well as a more uniform overall distribution.

4.4 ITS Analysis

After looking at the MFT, we turned our attention to the ITS. The first step was to look at the same variables as with the MFT. We would expect that the distributions be uniform in some sense, and unlike the MFT they should be cylindrically symmetric. The first issue, though, is making sure the data we're looking at is only coming from the ITS. The table that contains tracks from the central region is `aod::Tracks`, but since this has tracks as determined by all the detectors in the central region, such as the TPC and TOF, we had to narrow things down so we could study the ITS alone. In order to make sure only ITS tracks are considered, we used the table `aod::TracksIU` (Innermost Update) alongside the flag `hasITS`. The table returns the tracks determined at the innermost update, which is not always the ITS, so we use the `hasITS` flag to make sure that it is.

This method is not foolproof but in the rigidly structured data model it was the best we could do without reworking the entire thing or undertaking the reconstruction ourselves.

Figure 4.6 shows some histograms of the data before using the `hasITS` flag. The red histograms are without any modification and the green is choosing all tracks that are NOT seen in the ITS. These should be discussed in conjunction with the histograms of just the ITS tracks, which are presented in figure 4.7.

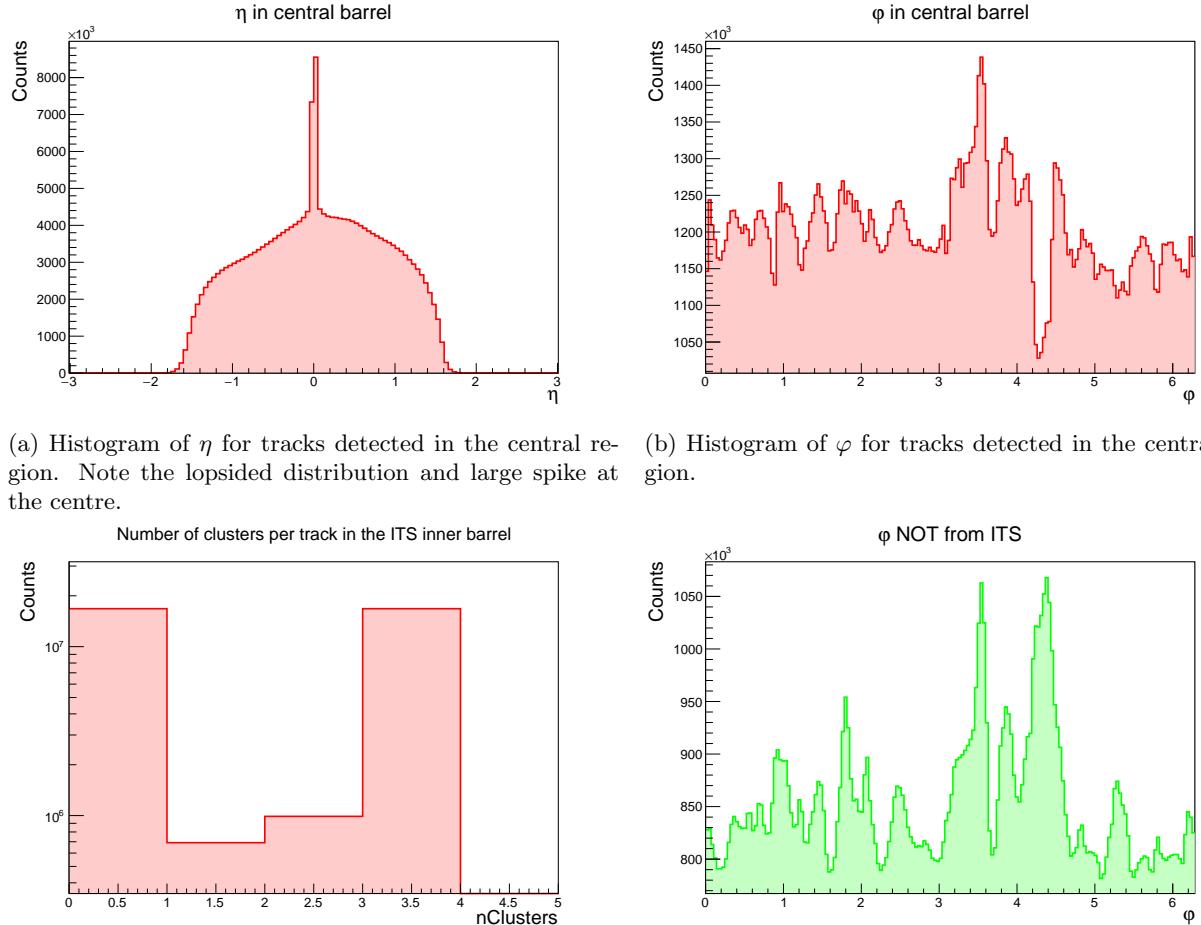
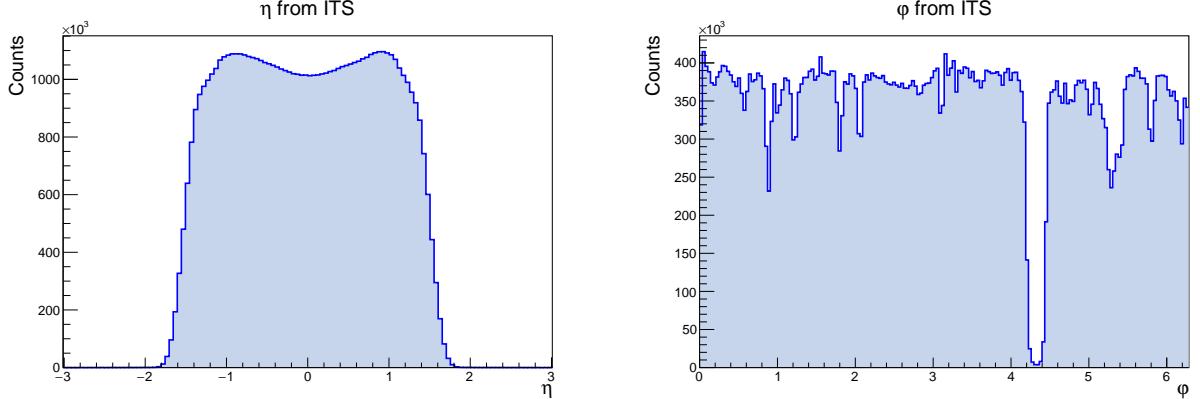


Figure 4.6: Some histograms of kinematic variables of tracks detected in the central region. This is data for the tracks at their innermost update, but that does not necessarily mean the tracks are only from the ITS. Uniform and symmetric distributions (aside from the number of clusters) are expected but not seen.

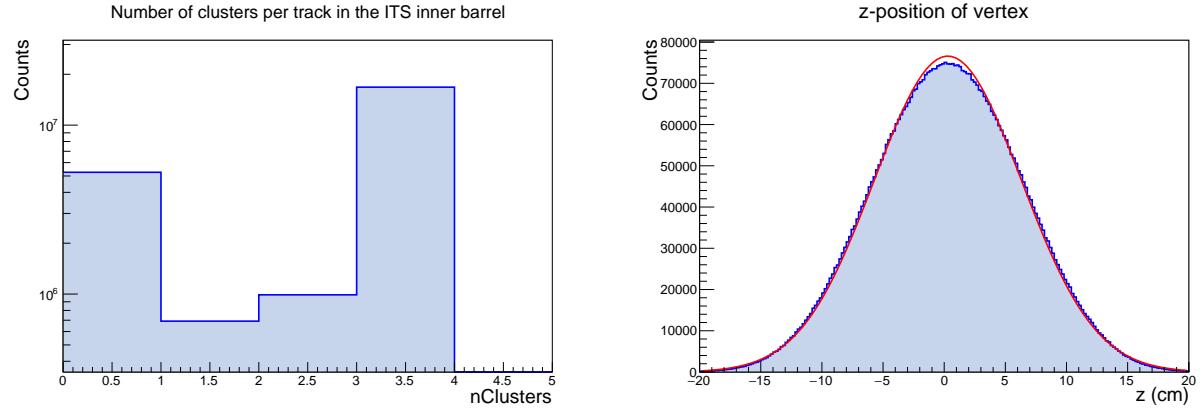
Upon seeing the ITS tracks in figure 4.7, a few questions arise. Firstly, we see how the η distribution is remarkably uniform and symmetric, especially considering the form of figure 4.6a. However, there is clearly a dip in the middle, which contradicts our reasoning for the lopsided distribution in the MFT as we would expect a distribution peaked at the centre and slowly decreasing as it goes outwards. This might be explained by the number of tracks at that pseudorapidity being so large that pile-up or overloading occurs, leading to missed tracks. Figure 4.6a supports our expectations

as it is sharply peaked at the centre. This requires more investigation alongside the read-out architecture.



(a) Histogram of η for tracks detected in the ITS. Note the uniform and symmetric distribution, but with a valley in the centre.

(b) Histogram of φ for tracks detected in the ITS. Note how the distribution is overall much more uniform, but with a large valley around 4.3.



(c) Histogram of number of clusters in the ITS inner barrel that contribute to a given track. The number of tracks with 0 clusters is far fewer than in figure 4.6c

(d) Histogram of the z -position of the collision vertex. A gaussian function is fit to the data, with $\mu = 0.2951$, $\sigma = 6.0205$. Note that this is not the position as determined by the ITS alone as that information is not in the data model at present. Each entry is a collision, not a track.

Figure 4.7: Histograms of kinematic variables of tracks detected by the ITS as well as of collisions.

The φ distribution in figure 4.7b looks much more uniform, at least on the whole, than in figure 4.6b. This is made more obvious when taking figure 4.6d into account as there appears to be no pattern or symmetry in the φ distribution of tracks when not detected in the ITS. No reason for this is apparent to us so this requires a more in-depth investigation. The valley in figure 4.7b at about 4.3, and smaller ones in other places, is certainly a concern. Figure 4.8 helps to elucidate the situation as we can see there is no η dependence on the large gap, so we are inclined to say that a whole strip of detectors in the ITS, running along the z -axis, was not functioning properly. This is reasonable as the electronics in the ITS group pixel chips together in that direction. Other smaller gaps can also be seen, spanning short ranges of η , which would imply short strips of pixel chips weren't working properly.

Looking at figure 4.7c in comparison to figure 4.6c, we see a decrease in the number of tracks with no clusters in the inner barrel. This is clearly as expected. Lastly, the z -position of the

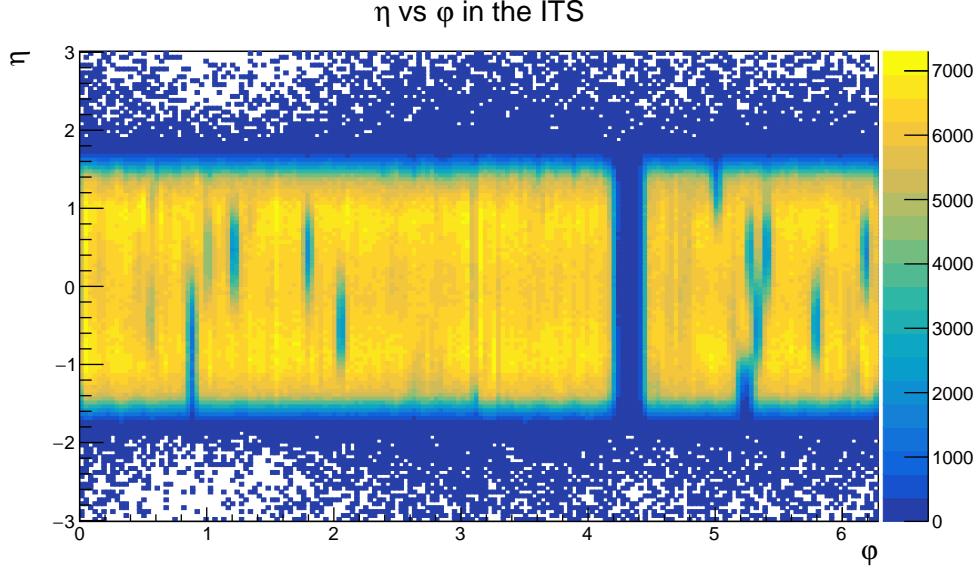
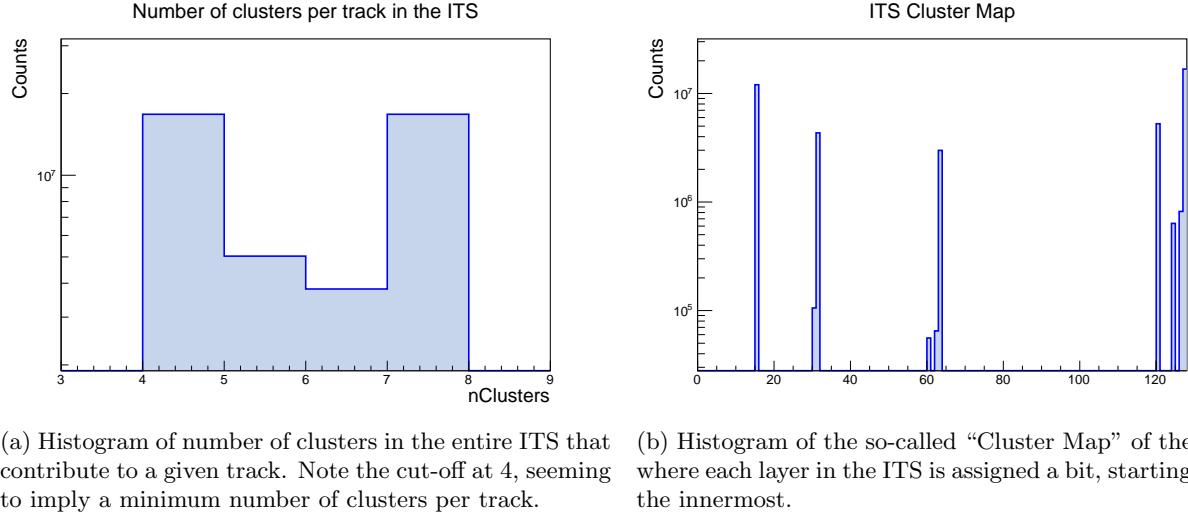


Figure 4.8: Histogram of η and φ for tracks detected in the ITS. The valley at $\varphi \approx 4.3$ can clearly be seen, with no specific η region being affected. Note a few other smaller strips of lower counts.

collision vertex is shown in figure 4.7d with a gaussian function fit to it. Important to note here is that each entry in the histogram is not a track, as it has been before, but rather a collision in the `aod::Collisions` table. While the ITS is the main detector that determines the vertex position, there is little information about what information was used to determine the vertex so little can be said about how the ITS performs in this regard. Nevertheless, this is a useful plot to look at to make sure that collisions are happening both as close to the IP as possible but also symmetrically around the IP. Both these conditions are met in this case, within acceptable tolerance ($\mu = 0.2951$, $\sigma = 6.0205$).

The last histograms to interrogate are those corresponding to the clusters contributing to tracks in the ITS, shown in figure 4.9. The left is easy to interpret, showing the number of ITS clusters used in each track. Interestingly it appears to have a minimum of 4 clusters per track. If these tracks are from the TPC-ITS track matching regime, this doesn't make sense, but if they are standalone tracks then we may have discovered something about the method used.

On the right of figure 4.9 is the “Cluster Map” of the ITS. The column in the data model contains an 8 bit unsigned integer with each bit assigned to a layer, excluding the last bit, for each track. If a track has a cluster in a specific layer, the bit is set to 1, otherwise it is set to 0. Since integers have unique representations in binary, we can see exactly which layers contributed to a given track simply by decoding the integer. As seen in figure 4.9b, the lowest is 15 meaning the innermost 4 layers had clusters contributing, and the highest is 127 meaning all layers had clusters contributing. In between those is all combinations of the 7 bits, but only those with 4 or more clusters, as expected from figure 4.9a.



(a) Histogram of number of clusters in the entire ITS that contribute to a given track. Note the cut-off at 4, seeming to imply a minimum number of clusters per track.

(b) Histogram of the so-called “Cluster Map” of the ITS, where each layer in the ITS is assigned a bit, starting from the innermost.

Figure 4.9: Additional histograms with information about the ITS clusters used in track finding.

5 Discussion and Recommendations

Our methods of analysis for the most part came down to interpreting histograms of kinematic variables through the lens of assumed isotropic emission of particles. The shape of the coordinate system and detector geometry had to be considered in order to explain some things and there was a lot of guesswork as to the configuration of the detectors and reconstruction process/analysis framework. In this section we present some improvements that can be made to this process.

There were many times during our analysis where we came up against a limitation in the analysis software that restricted our ability to see the details we were after. This really comes down to the fact that the types of analysis that O2 was (and still is being) built for are not the type that we were looking to do. Given more time to attain a greater understanding of the inner workings of particularly the steps from CTFs to AODs, we might have been able to delve into the belly of the beast and find the data ourselves. Since we were limited by our time this year, the best we can do is recommend some modifications to the AOD data model that would lend themselves to these analyses.

We recommend introducing more detail about the tracks in the MFT, such as the position of hits in each disk, or even a simple “Cluster Map”-style column in the `aod::MFTTracks` table to be able to see which disks or detector planes contributed to the track. At time of writing, there is no way to view the vertex position data as determined by the ITS or the MFT alone. Adding this capability would allow for an investigation into the improvement of the new ITS and MFT in that sector, which we were unable to do.

Outside the realm of modifying O2 and the data model, there is more that could be done to properly verify if the distributions we see for, say η and φ , are correct. We have attempted to explain them from a physical motivation but comparison to simulated data would prove the most robust way to check that there are no large systematic errors. In the interest of determining how the MFT has improved the tracking capabilities in the forward region, the old techniques used in Runs 1 and 2 using the MCH could be implemented on new data and compared to the new technique on the same data. This would require investigation of the analysis framework at a much more involved level than was possible in this report but could lead to some interesting results. Once again simulation could aid in this endeavour.

Lastly, we were left with many questions about the details of the track reconstruction, read-out architecture, and columns in the data model such as the z variable in `aod::MFTTracks`. This effectively comes down to the documentation surrounding these detectors and analysis framework, so a concerted effort to improve that documentation would be greatly appreciated by people new to the ALICE group and to these detectors.

6 Conclusion

The advent of Run 3 at the LHC, with its increased energy and luminosity, brought a number of new challenges that required the ALICE detectors and analysis framework to be upgraded. Among the detector upgrades were two new silicon tracking detectors; the ITS and MFT. In this report we used the new O2 analysis framework to look at distributions of variables relating to particle tracks in either the ITS or MFT, using pilot beam data from October 2021. The intention was both to learn the new framework and to determine if the new detectors are performing as expected.

Before any analysis could be done, the new framework had to be learned. Due to its need to be efficient and lightweight on hard disk and memory usage, it is very particular about how things must be structured. Learning O2 took up the majority of the time spent on this project and the basics are presented in section 3.

Simulation of data for comparison was outside the scope of this project so the notion of an “expected performance” had to come from an understanding of the physical design of the detectors as well as an interpretation of the read-out architecture and track reconstruction process as outlined in the Technical Design Reports of the two detectors. Isotropic emission of particles from the IP was assumed which led to expectations of smooth, if not uniform, distributions in most cases. Where this was not seen and the geometry of the sensitive material of the detectors couldn’t explain the discrepancy, specific areas of research into the detectors were recommended. This analysis is presented in section 4.

Due to time limitations, this report could only give a qualitative review of the performance of the new detectors so we present recommendations for the direction of further investigation in section 5 if a more quantitative perspective is desired.

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Appendix

A Downloading data from the GRID

```
1  # The directory in alimonitor where you want to get the data from. Should contain a load of
2  ↵ numbered directories
3  # For example, /alice/data/2021/OCT/505548/AOD. Note that it's AOD not A02D
4  sourceMotherDir=/alice/data/path/to/AOD
5  nFiles=$1
6
7  # This is the directory on your local machine where you want to store your data
8  targetMotherDir=/path/to/alice/data/${sourceMotherDir}
9  mkdir -p ${targetMotherDir}
10
11 for ((i=1; i<=nFiles; i++))
12 do
13     if (( i < 10 )); then
14         pref=00
15     fi
16     if (( i > 9 )); then
17         pref=0
18     fi
19     iDir=${pref}${i}
20     iSourceDir=${sourceMotherDir}/${iDir}
21     iTarGetDir=${targetMotherDir}/${iDir}
22     mkdir -p ${iTarGetDir}
23
24     echo "copying from ${iSourceDir} to ${iTarGetDir}"
25
26     alien_cp -retry 5 ${iSourceDir}/A02D.root file:///${iTarGetDir}/A02D.root
done
```

B Additional Plots

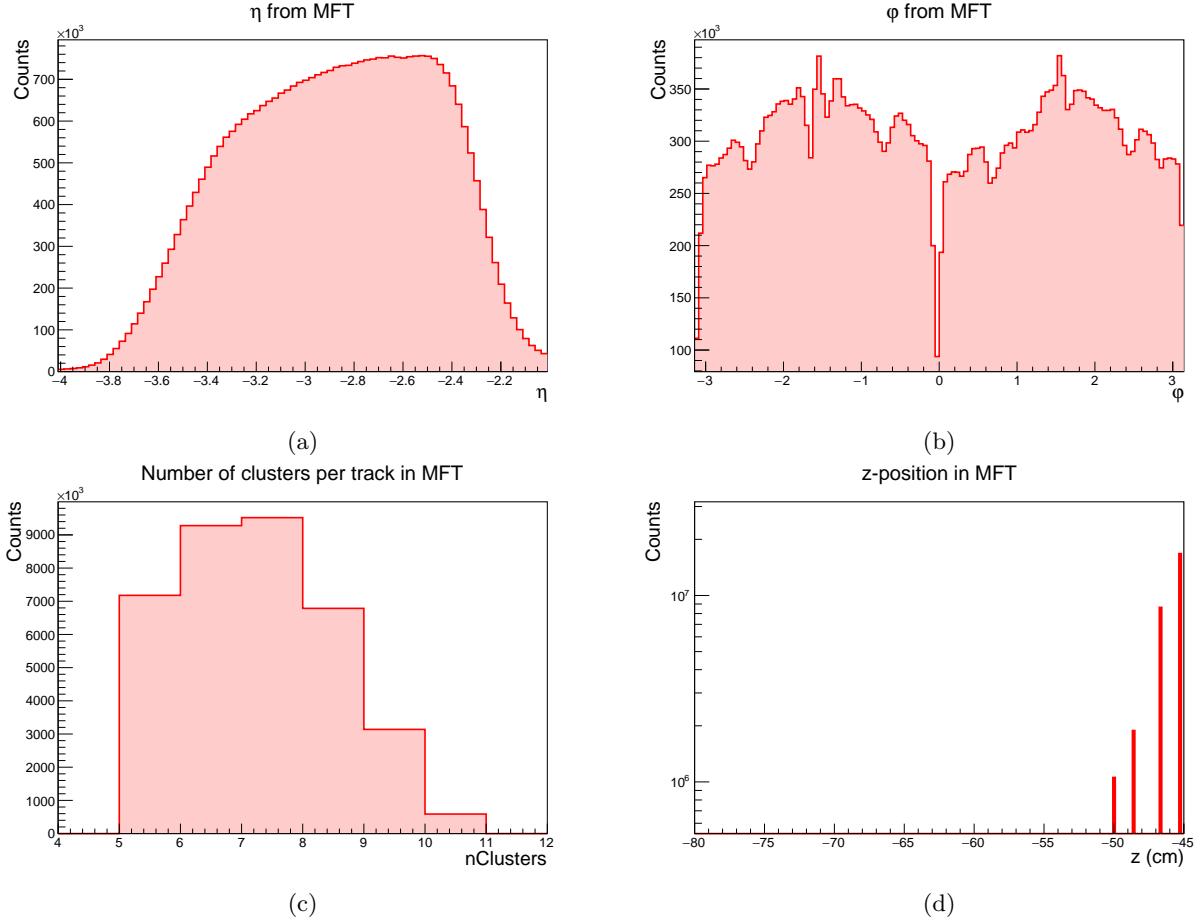


Figure B.1: Histograms of kinematic variables for tracks in the MFT from reconstruction pass 4. Presented here for comparison to figure 4.1. Note that other than η there is no significant difference between these and those from pass 3.

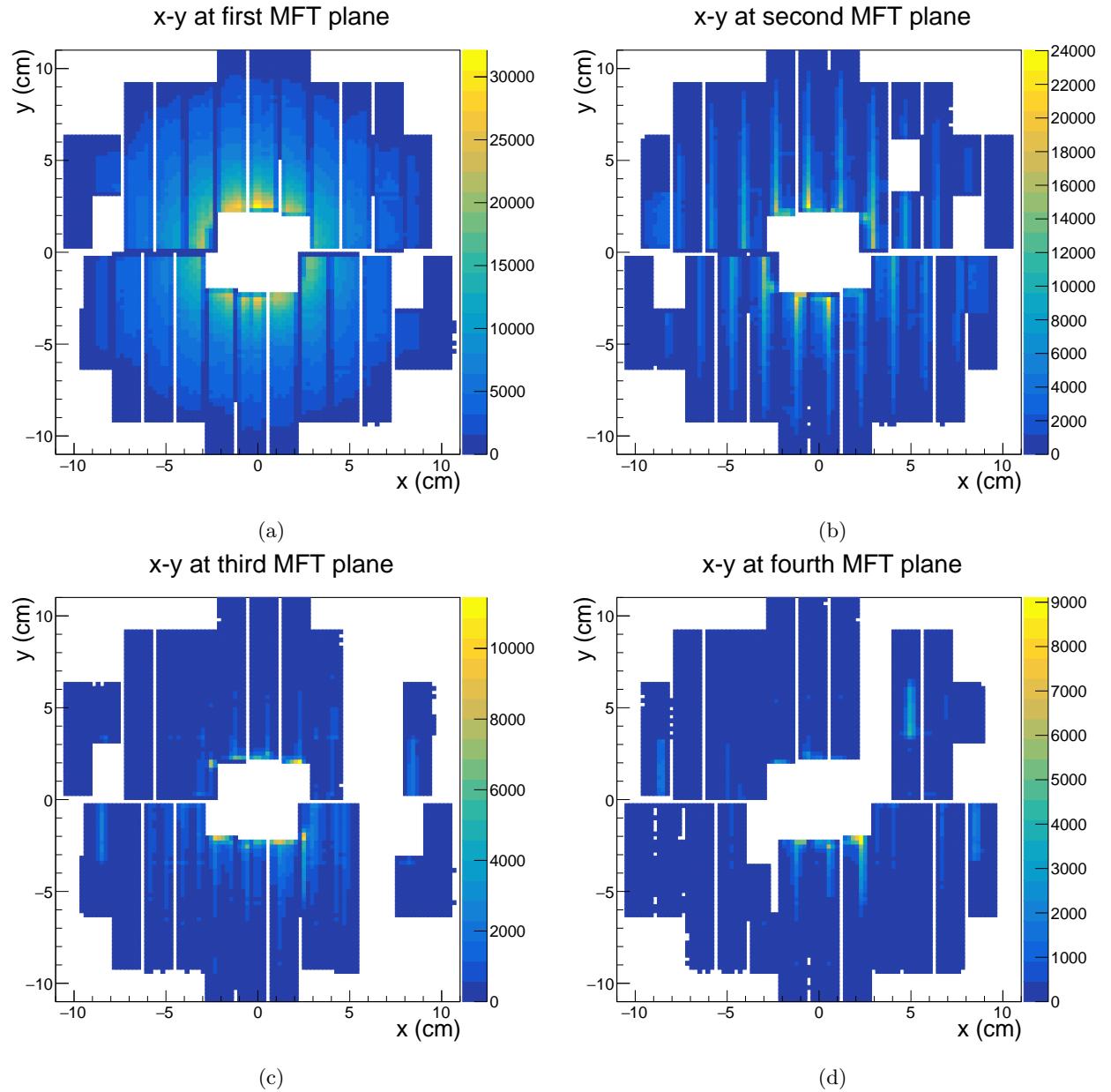


Figure B.2: x - y histograms for MFT tracks from reconstruction pass 4. Once again there is not much difference compared to pass 3 other than a general trend towards more uniform data.