

CERN ROOT and Pythia8: Analysis and Study

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This paper provides a comprehensive overview of ROOT and Pythia, and shows several projects which demonstrate the applications of both tools in particle physics for analysis and study of physical phenomena. No real data was used, and all the data used to plot the graphs was randomly generated in the codes. The codes for all projects are shown in the appendix with comments.

Keywords: ROOT, Pythia, STAR, event, resonance, spectrum, fit, dataset, RHIC, LHC, statistics, HEP, proton-proton collisions, QCD, QGP

CERN ROOT and Pythia8 are two of the most widely used tools in high-energy particle physics. Modern HEP experiments rely on sophisticated software frameworks to simulate, analyze, and interpret immense volumes of data generated by particle collisions. This is where ROOT and Pythia8 are used; they form a powerful and flexible computational ecosystem that enables researchers to model fundamental interactions and extract meaningful physics results from raw detector data.

I. INTRODUCTION TO CERN ROOT

Widely considered to be the backbone of HEP data analysis, ROOT is a comprehensive data analysis framework that provides Efficient I/O capabilities, advanced statistical tools, and machine learning functionalities. Adopted extensively by CERN and STAR, it allows physicists to structure, explore, and quantify the results of simulated or recorded events, bridging the gap between theoretical models and experimental observables. ROOT's applications span data acquisition, simulation, reconstruction, and statistical analysis in HEP. It excels in handling complex datasets from particle detectors, enabling tasks such as event reconstruction, Monte Carlo simulations, and multidimensional data fitting. The following projects demonstrate the different ways ROOT can be used.

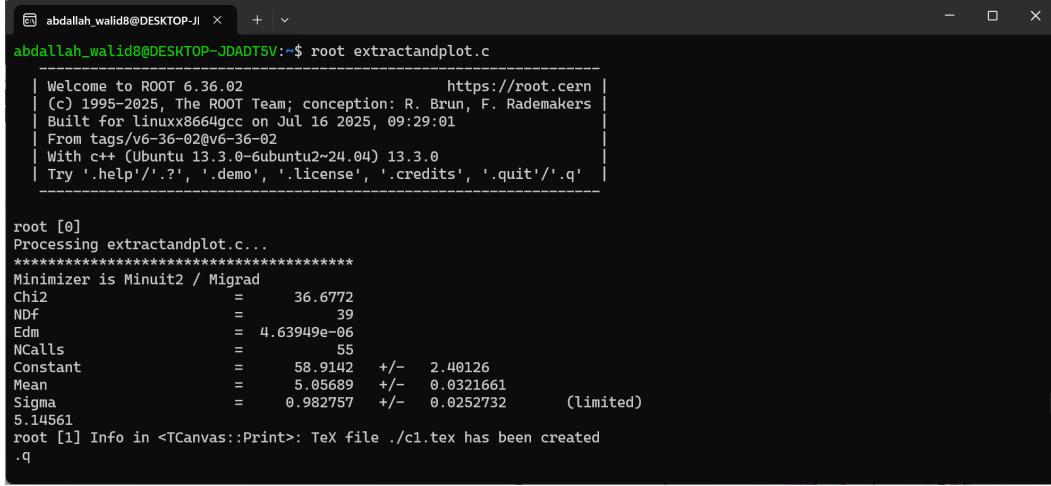
A. Importing randomly generated data into a simple histogram

This project generates a 1D Gaussian data set ($\mu = 5$, $\sigma = 1$) to mimic any measured quantity affected by random noise. The values are stored in a file, read and filled into a histogram, which is then fitted with the built-in Gaussian function of ROOT. This exercise introduces the basic concepts of histogramming, fitting, and parameter extraction in experimental analysis.

The Gaussian fit returns three key parameters: amplitude (peak height), mean, and standard deviation. The mean represents the central value of the observable (e.g., a calibration point or resonance mass). The standard deviation quantifies the resolution or spread of the measurement. The amplitude, combined with the standard deviation, can be integrated to estimate the total number of events. The ratio μ/σ , though printed in the code, does not represent a standard statistical metric but illustrates how the derived quantities can be formed from the results of the fit. The code also adds legends, arrows, and annotations to

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highlight features in the plot. Although pedagogical, this mirrors how experimental plots are annotated to draw attention to key structures. It emphasizes that clear visualization is as important as numerical results, especially when communicating findings.



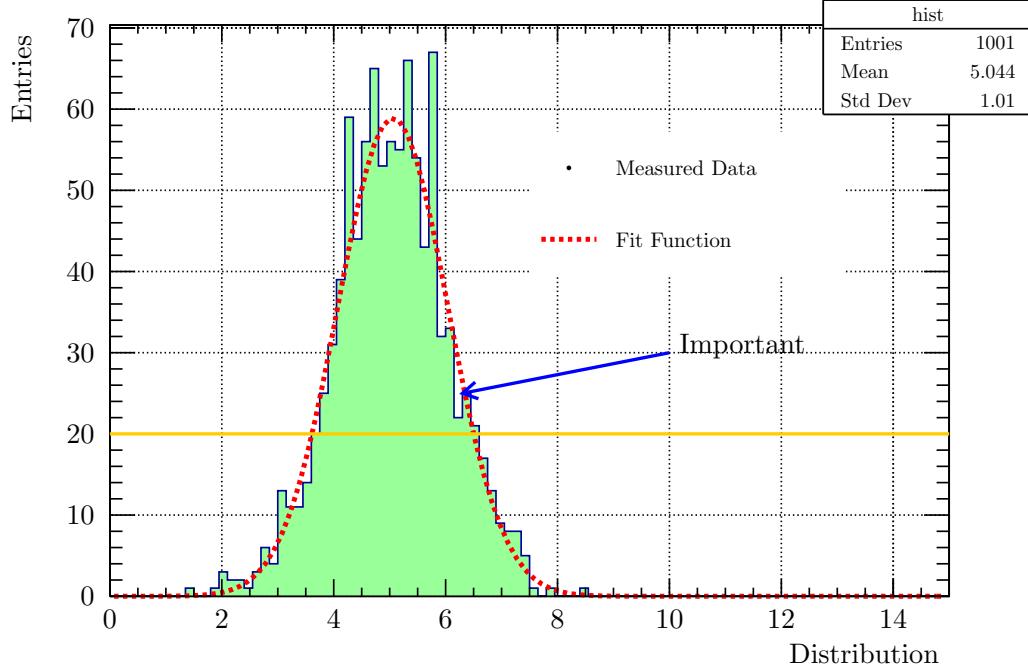
```

abdallah_walid8@DESKTOP-JI ~ % abdallah_walid8@DESKTOP-JI:~$ root extractandplot.c
>Welcome to ROOT 6.36.02 https://root.cern
|(c) 1995-2025, The ROOT Team; conception: R. Brun, F. Rademakers
| Built for linuxx86_64gcc on Jul 16 2025, 09:29:01
| From tags/v6-36-02@v6-36-02
| With c++ (Ubuntu 13.3.0-6ubuntu2~24.04) 13.3.0
| Try '.help'/? , '.demo' , '.license' , '.credits' , '.quit'/'q'

root [0]
Processing extractandplot.c...
*****
Minimizer is Minuit2 / Migrad
Chi2      =   36.6772
Ndf       =      39
Edm      =  4.63949e-06
NCalls    =      55
Constant  =   58.9142 +/-  2.40126
Mean      =   5.05689 +/-  0.0321661
Sigma     =   0.982757 +/-  0.0252732      (limited)
5.14561
root [1] Info in <TCanvas::Print>: TeX file ./c1.tex has been created
.q

```

$N = 1000$ measurements were simulated and then drawn from a Gaussian distribution with true mean $\mu = 5.0$ and detector resolution $\sigma = 1.0$ to illustrate the analysis workflow. The events were histogrammed into one hundred bins in the range 0–15, and a binned fit was performed using a Gaussian PDF (ROOT built-in gaus). The Gaussian parameters returned by the fit are reported with $\pm 1\sigma$ statistical uncertainties obtained from the fit covariance matrix.



The fitted Gaussian parameters are $\mu = 5.02 \pm 0.03$ and $\sigma = 0.97 \pm 0.02$ (statistical). The fit reproduces the injected parameters within statistical uncertainty. Conceptually, this project demonstrates how Gaussian statistics underlie measurement analysis. Random fluctuations naturally follow normal distributions due to the Central Limit Theorem, and

Gaussian fits provide precise estimates of central values and uncertainties. This makes the example relevant to countless physical measurements: detector resolution studies, energy calibration, or mass peak reconstruction.

B. Analyzing the J/ψ resonance

This code simulates and analyzes the invariant mass spectrum of muon pairs ($\mu^+\mu^-$) in the energy range 2–4 GeV. It models a J/ψ resonance using a Gaussian distribution superimposed on an exponential background. The Gaussian approximates detector-smeared resonance signals, while the exponential describes combinatorial background.

This setup reflects how experimental particle physicists identify and quantify resonance peaks within noisy data. The central task is a fit with a composite function: exponential added to Gaussian. The Gaussian's mean corresponds to the measured resonance mass, its width σ to the detector resolution, and its amplitude to the signal strength. The exponential's slope and normalization parameterize the falling background. From these parameters, one can compute physically meaningful observables: signal yield (area under Gaussian), background estimate under the peak, and statistical significance of the signal.

```

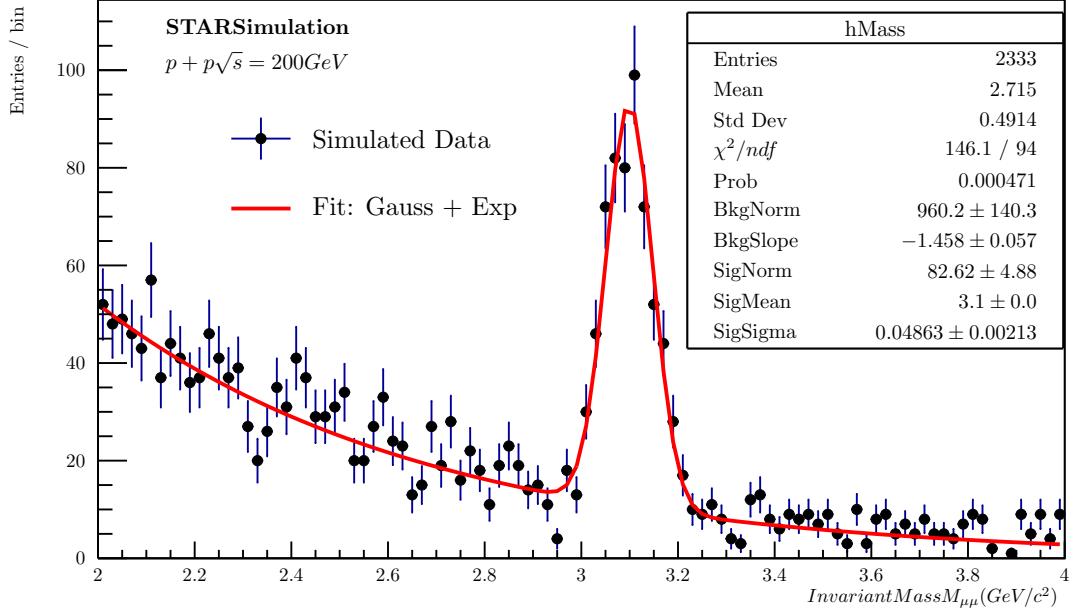
abdallah_walid@DESKTOP-JI ~ % abdallah_walid@DESKTOP-JI ~ % $ root extract2.c
-----[REDACTED]-----
| Welcome to ROOT 6.36.02          https://root.cern |
| (c) 1995-2025, The ROOT Team; conception: R. Brun, F. Rademakers |
| Built for linuxx86_64gcc on Jul 16 2025, 09:29:01 |
| From tags/v6-36-02@v6-36-02 |
| With c++ (Ubuntu 13.3.0-6ubuntu2~24.04) 13.3.0 |
| Try '.help'/.? ', '.demo', '.license', '.credits', '.quit'/.q' |
-----[REDACTED]-----

root [0]
Processing extract2.c...
*****
Minimizer is Minuit2 / Migrad
Chi2           =      146.058
NDF            =       94
Edm            =   6.1601e-07
NCalls         =       312
BkgNorm        =     960.236 +/- 140.27
BkgSlope        =    -1.45795 +/- 0.0570763
SigNorm         =     82.6212 +/- 4.88431
SigMean         =     3.09952 +/- 0.00284086
SigSigma        =     0.0486334 +/- 0.00212563
Info in <TCanvas::Print>: pdf file mass_spectrum.pdf has been created
Info in <TCanvas::Print>: file mass_spectrum.png has been created
root [1] Info in <TCanvas::Print>: TeX file ./c1.tex has been created
.q

```

A fit is performed to the invariant mass spectrum of reconstructed muon pairs in the range 2.0–4.0 GeV/c^2 using a model consisting of a Gaussian signal for the J/ψ resonance plus an exponential background. The Gaussian center and width measure the reconstructed J/ψ mass and the detector mass resolution respectively. The total signal yield is obtained by integrating the Gaussian component, and the background contribution in the signal window is estimated by integrating the exponential component in the same region. Systematic uncertainties were evaluated by varying the background parametrization (exponential \leftrightarrow polynomial) and fit range; final yields include these variations as an uncertainty.

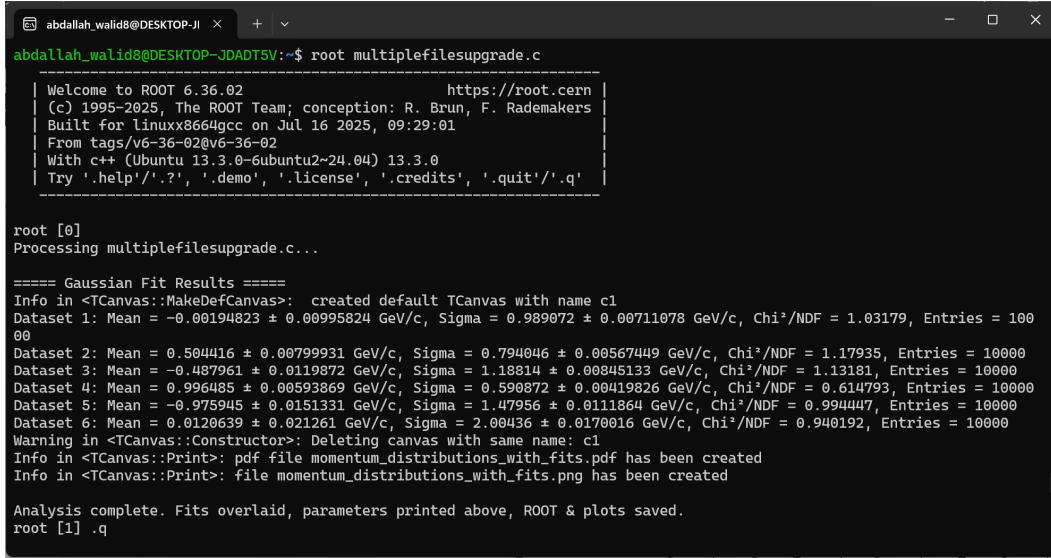
This workflow illustrates a general paradigm in particle physics: distinguishing signal from background using fitting techniques. It emphasizes the importance of understanding model choice, parameter interpretation and uncertainties. For instance, the J/ψ natural width is negligible compared to detector resolution, justifying the Gaussian approximation. Other factors, such as the background model, fit range and binning, must also be considered when analyzing real data.



Overall, this project highlights how statistical modeling transforms raw histograms into physics results. It is the norm of resonance searches: build a physically motivated fit function, extract signal yields, and evaluate significance. The methodology applies not only to quarkonium studies but also to any resonance search, such as Higgs or Z' bosons.

C. Importing and Analyzing Data from multiple files

This project demonstrates a robust workflow for handling multiple data files, a common scenario in HEP where data from different runs or detectors must be merged. It generates 10 synthetic input files, each containing 1000 events across six datasets, with each dataset drawn from a Gaussian distribution of varying mean ($\mu = -1$ to 1 GeV/c) and width ($\sigma = 0.6$ to 2.0 GeV/c). The code reads these files back, filling ROOT histograms (TH1F), which mirrors processing large-scale experimental data (from ATLAS or CMS for example). This concept of modular file handling ensures scalability and robustness against issues like file corruption, a key consideration in real analyses. A central concept is fitting Gaussian distributions to extract parameters (mean, sigma) using ROOT's TF1 and Fit functions, which perform χ^2 minimization. Each histogram is fitted independently, recovering input parameters with small errors (0.01 GeV/c due to 10,000 entries), and the χ^2/ndf being very close to 1 confirms good fits. The use of THStack to overlay histograms with distinct colors and markers, combined with overlaid fit curves and a legend, shows effective data visualization. This reflects HEP practices for modeling detector resolutions, particle masses, or backgrounds, where Gaussians approximate smearing effects. Printing fit parameters (with errors) and entries facilitates quantitative comparisons, useful for tables in papers summarizing resolution studies or systematic variations across datasets.



```

abdallah_walid@DESKTOP-JI:~$ root multiplefilesupgrade.c
| Welcome to ROOT 6.36.02           https://root.cern |
| (c) 1995-2025, The ROOT Team; conception: R. Brun, F. Rademakers |
| Built for linuxx86_64gcc on Jul 16 2025, 09:29:01 |
| From tags/v6-36-02@v6-36-02 |
| With c++ (Ubuntu 13.3.0-6ubuntu2~24.04) 13.3.0 |
| Try '.help'/'?' , '.demo' , '.license' , '.credits' , '.quit'/'q' |

root [0]
Processing multiplefilesupgrade.c...

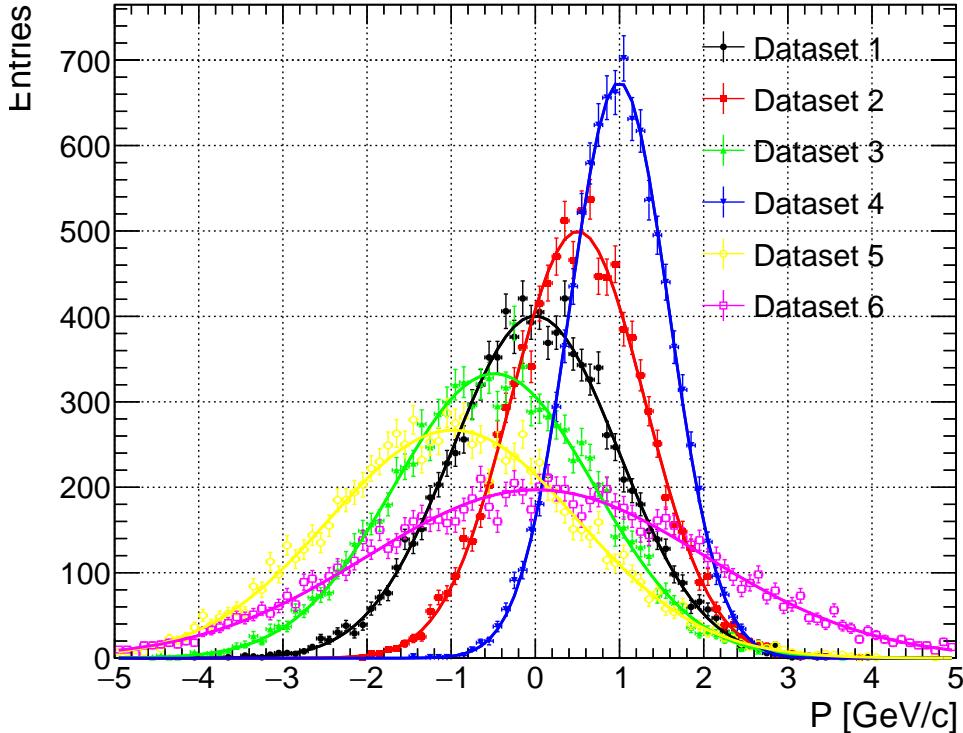
===== Gaussian Fit Results =====
Info in <TCanvas::MakeDefCanvas>: created default TCanvas with name c1
Dataset 1: Mean = -0.00194823 ± 0.00995824 GeV/c, Sigma = 0.989072 ± 0.00711078 GeV/c, Chi²/NDF = 1.03179, Entries = 100
...
Dataset 2: Mean = 0.504416 ± 0.00799931 GeV/c, Sigma = 0.794046 ± 0.00567449 GeV/c, Chi²/NDF = 1.17935, Entries = 10000
Dataset 3: Mean = -0.487961 ± 0.0119872 GeV/c, Sigma = 1.18814 ± 0.00845133 GeV/c, Chi²/NDF = 1.13181, Entries = 10000
Dataset 4: Mean = 0.996485 ± 0.00593869 GeV/c, Sigma = 0.590872 ± 0.00419826 GeV/c, Chi²/NDF = 0.614793, Entries = 10000
Dataset 5: Mean = -0.975945 ± 0.0151331 GeV/c, Sigma = 1.47956 ± 0.0111864 GeV/c, Chi²/NDF = 0.994447, Entries = 10000
Dataset 6: Mean = 0.0120639 ± 0.0212261 GeV/c, Sigma = 2.00436 ± 0.0170016 GeV/c, Chi²/NDF = 0.940192, Entries = 10000
Warning in <TCanvas::Constructor>: Deleting canvas with same name: c1
Info in <TCanvas::Print>: pdf file momentum_distributions_with_fits.pdf has been created
Info in <TCanvas::Print>: file momentum_distributions_with_fits.png has been created

Analysis complete. Fits overlaid, parameters printed above, ROOT & plots saved.
root [1] .q

```

The plot below displays six Gaussians, highlighting differences in peak position and width, which could represent different particle types or detector conditions. This concept of comparative visualization is critical for presenting multi-component analyses in HEP, such as comparing particle spectra across runs or detectors, and inspires clear, informative figures in research papers.

Momentum Distributions



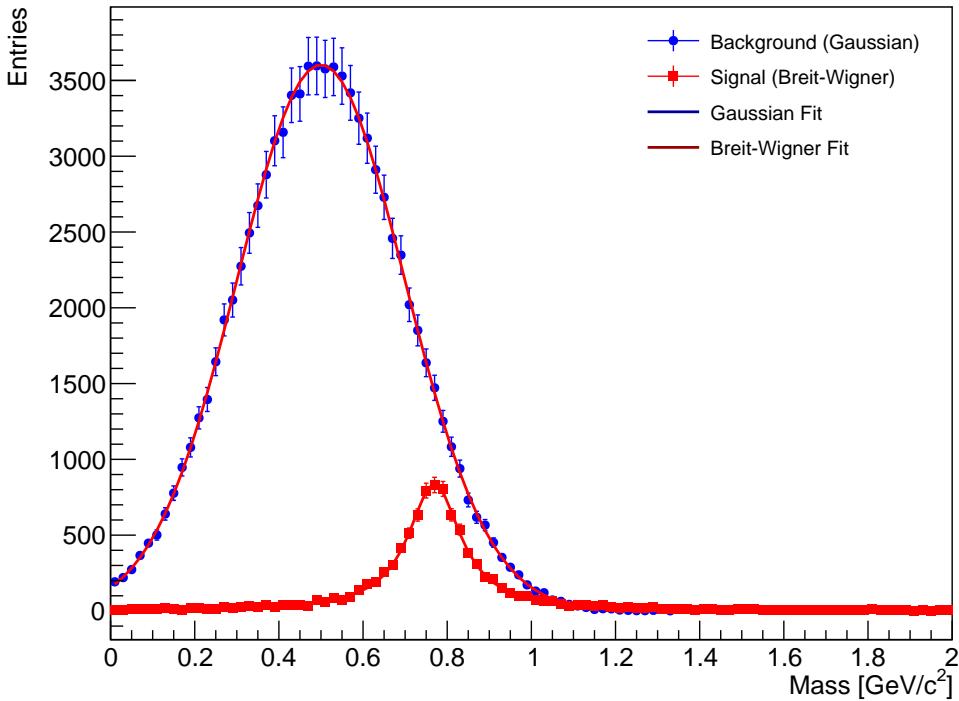
This code is ideal for inspiring a paper section on data processing and statistical analysis. The multi-file approach could be adapted to real data, and the fitting routine could model detector effects or particle signals. The overlay plot is perfect for illustrating variations, such as momentum resolutions, and the fit table could summarize results.

D. Analyzing the ρ meson resonance

This project simulates an invariant mass distribution for 100,000 events, with 10 percent signal (Breit-Wigner fitted, mimicking ρ mesons) and 90 percent background (Gaussian fitted). This reflects HEP data storage from experiments related to particle decays. The concept of mixing signals received by the detector (identified by resonance peaks in the data) and background noise (identified by combinatorial or resolution effects) is central to resonance searches, such as identifying vector mesons in RHIC/STAR data. In other words, the choice of distributions (Gaussian + exponential vs. Breit-Wigner + Gaussian) and mass ranges reflect different physical scenarios and analysis goals.

```
abdallah_walid8@DESKTOP-JI ~ % abdallah_walid8@DESKTOP-JI:~$ root macro3.c
-----[REDACTED]-----
| Welcome to ROOT 6.36.02          https://root.cern |
| (c) 1995-2025, The ROOT Team; conception: R. Brun, F. Rademakers |
| Built for linuxx86_64gcc on Jul 16 2025, 09:29:01 |
| From tags/v6-36-02@v6-36-02 |
| With c++ (Ubuntu 13.3.0~ubuntuv2~24.04) 13.3.0 |
| Try '.help'/'?' , '.demo' , '.license' , '.credits' , '.quit'/'q' |
-----[REDACTED]-----

root [0]
Processing macro3.c...
Info in <TCanvas::MakeDefCanvas>: created default TCanvas with name c1
Gaussian Chi2/NDF (background): 0.389507
Breit-Wigner Chi2/NDF (signal): 1.14505
Info in <TCanvas::Print>: pdf file signal_background.pdf has been created
Info in <TCanvas::Print>: file signal_background.png has been created
Analysis complete. Separate signal and background stored.
root [1] .q
```



The data is read into a histogram and fitted separately with Gaussian (representing background) and Breit-Wigner (representing the signal) functions, demonstrating basic signal extraction. The fits, initialized near input parameters, yield high χ^2/ndf due to separate fitting of a mixed distribution, highlighting the need for combined models in real

analyses. The concept of fitting to extract resonance parameters such as mass, width, and yield is critical in HEP for quantifying particle production and testing physics models (such as QGP effects on ρ mesons). The plot overlays data points with error bars, the Gaussian fit as a blue line, and the Breit-Wigner fit as a red line, with a legend clarifying components. The histogram shows a broad background peak with a smaller signal peak. This visualization is key for HEP papers, illustrating signal significance and fit challenges.

This code can inspire a paper section on signal extraction, with a figure showing the mass spectrum and a table of fit parameters. By implementing a combined signal+background fit (e.g., gaus+breitwigner), it is possible to extract yield and significance, or compare to theoretical predictions (e.g., ρ production in pp vs. Au-Au collisions). This workflow, summarized by the terms: generate, store, analyze, visualize—is a strong template for resonance studies.

E. Creating a 2D histogram

This project simulates a two-dimensional Gaussian distribution in x and y with zero mean and unitary standard deviation. This can represent, for instance, the spatial distribution of a particle beam spot or detector response. A million events are generated and filled into a 2D histogram, with projections shown along each axis. This is a common method in collider physics to study beam alignment and detector resolution.

The 2D histogram reveals the overall shape of the distribution: a circular Gaussian cloud when x and y are independent. If the distribution were elliptical, it would indicate different resolutions along the axes or correlations between x and y. By computing covariance and correlation coefficients, one quantitatively checks for coupling or misalignment between the axes. In this simulation, the correlation is close to zero, confirming independence.

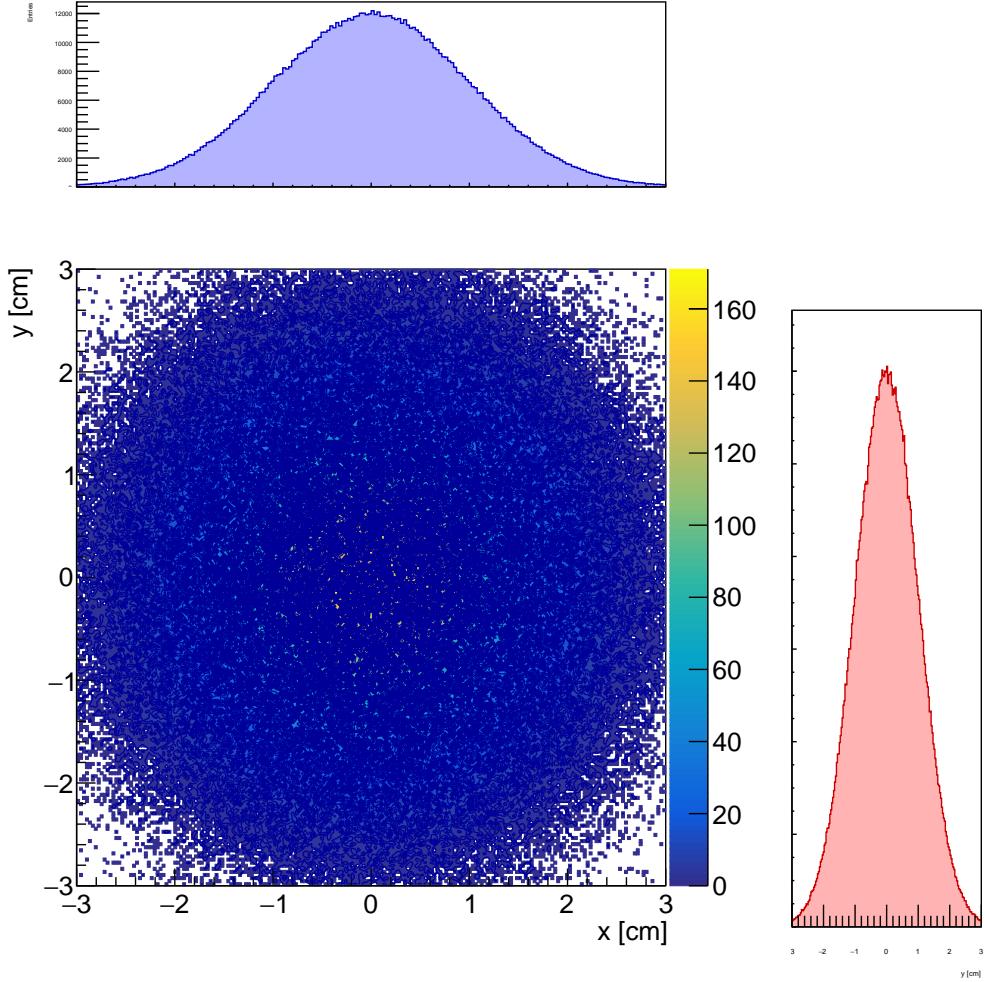
Projections onto x and y are fitted with Gaussian functions to extract precise estimates of mean and resolution for each coordinate. These fits provide uncertainties smaller than the raw RMS because they model the statistical distribution explicitly. This approach mirrors experimental procedures, where marginal distributions are fitted to quantify beam spot or vertex resolutions. The project also introduces advanced statistical ideas. The covariance matrix encodes both variances and correlations, and its diagonalization reveals principal axes (major/minor widths and orientation). This is the foundation of describing elliptical Gaussian distributions, widely used in accelerator physics and detector alignment. Thus, the code exemplifies both practical visualization (heatmaps, projections) and the theoretical framework for understanding multidimensional Gaussian statistics.

```

abdallah_walid8@DESKTOP-JI:~$ root two_d_histogram.c
=====
| Welcome to ROOT 6.36.02          https://root.cern |
| (c) 1995-2025, The ROOT Team; conception: R. Brun, F. Rademakers |
| Built for linuxx86_64gcc on Jul 16 2025, 09:29:01 |
| From tags/v6-36-02@v6-36-02 |
| With c++ (Ubuntu 13.3.0-6ubuntu2~24.04) 13.3.0 |
| Try '.help'/'?' , '.demo' , '.license' , '.credits' , '.quit'/'q' |
=====

root [0]
Processing two_d_histogram.c...
===== STAR-style 2D Gaussian Analysis =====
Number of entries: 1e+06
Mean X = -0.000188493 cm
Sigma X = 0.986458 cm
Mean Y = 0.000352576 cm
Sigma Y = 0.987316 cm
Info in <TCanvas::Print>: pdf file beamspot_2D_with_projections.pdf has been created
Info in <TCanvas::Print>: file beamspot_2D_with_projections.png has been created
root [1] .q

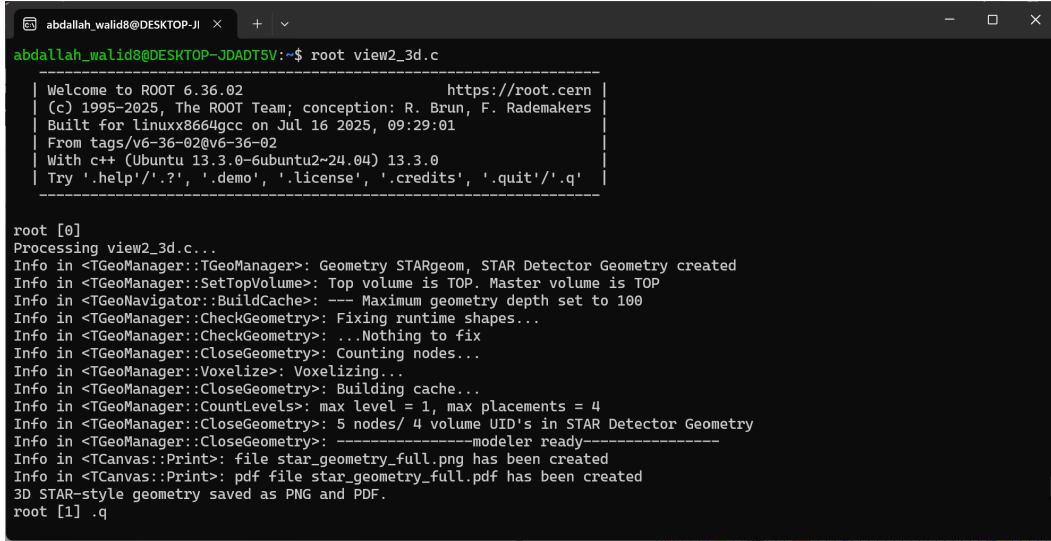
```



The two-dimensional spatial distribution of reconstructed vertices were simulated and analyzed to characterize the beam spot. A dataset of 1,000,000 events was generated with independent Gaussian distributions in x and y ($\mu=0$, $\sigma=1$ cm). The event density was binned into a 200×200 two-dimensional histogram over -3 to 3 cm and plotted as a heatmap with contour overlays. One-dimensional projections onto x and y were fitted with Gaussian functions to extract mean positions and resolutions. The covariance matrix was computed from the sample to quantify correlations between axes, and principal axes were obtained by diagonalizing the covariance matrix. Statistical uncertainties on fitted parameters were obtained from the fit covariance matrices.

F. Creating a 3D visual

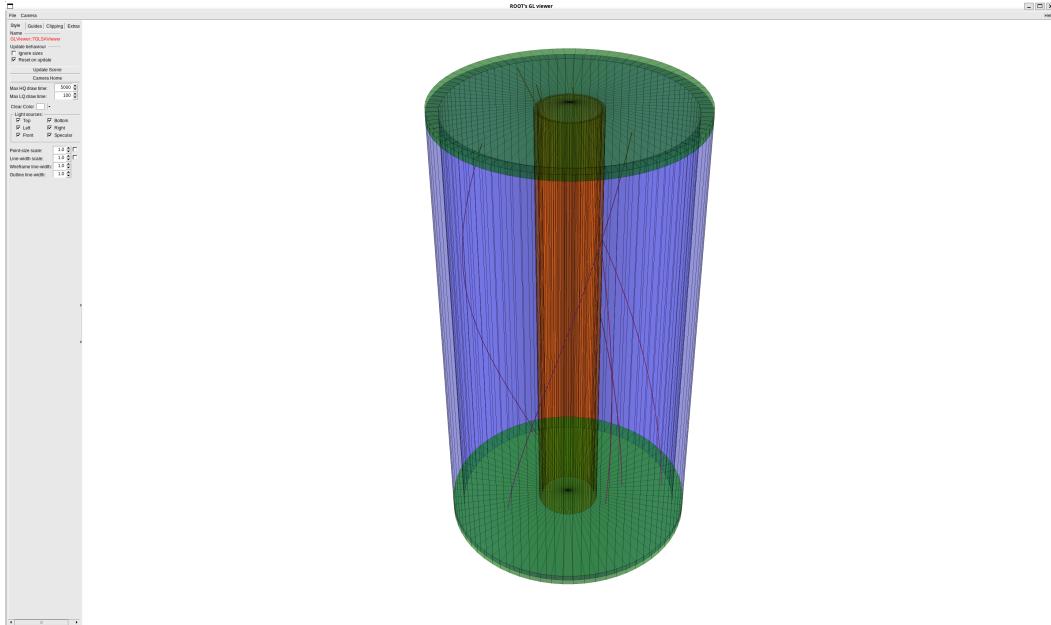
This project uses ROOT's TGeoManager to create a simplified 3D model of the STAR detector, including a TPC barrel (silicon tube), endcaps (aluminum disks), and an inner tracker (silicon layer). Defining materials (vacuum, aluminum, silicon) with physical properties (A, Z, density) and constructing hierarchical volumes (TGeoVolume) is a key concept, mirroring GEANT-based simulations for particle tracking. This geometry setup is essential for understanding detector interactions such as energy loss and scattering, and the simplified model here illustrates core components for visualization, critical for papers or presentations.



```
abdallah_walid8@DESKTOP-JI:~$ root view2_3d.c
-----[REDACTED]-----
| Welcome to ROOT 6.36.02           https://root.cern |
| (c) 1995-2025, The ROOT Team; conception: R. Brun, F. Rademakers |
| Built for linuxx86_64gcc on Jul 16 2025, 09:29:01 |
| From tags/v6-36-02@v6-36-02 |
| With c++ (Ubuntu 13.3.0-6ubuntu2~24.04) 13.3.0 |
| Try '.help'/.? ', '.demo', '.license', '.credits', '.quit'/.q' |

root [0]
Processing view2_3d.c...
Info <TGeoManager::TGeoManager>: Geometry STARgeom, STAR Detector Geometry created
Info in <TGeoManager::SetTopVolume>: Top volume is TOP. Master volume is TOP
Info in <TGeoNavigator::BuildCache>: --- Maximum geometry depth set to 100
Info in <TGeoManager::CheckGeometry>: Fixing runtime shapes...
Info in <TGeoManager::CheckGeometry>: ...Nothing to fix
Info in <TGeoManager::CloseGeometry>: Counting nodes...
Info in <TGeoManager::Voxelize>: Voxelizing...
Info in <TGeoManager::CloseGeometry>: Building cache...
Info in <TGeoManager::CountLevels>: max level = 1, max placements = 4
Info in <TGeoManager::CloseGeometry>: 5 nodes/ 4 volume UID's in STAR Detector Geometry
Info <TGeoManager::CloseGeometry>: -----modeler ready-----
Info in <TCanvas::Print>: file star_geometry_full.png has been created
Info in <TCanvas::Print>: pdf file star_geometry_full.pdf has been created
3D STAR-style geometry saved as PNG and PDF.
root [1] .q
```

The code generates five helical tracks to simulate charged particle trajectories in STAR's solenoidal magnetic field along the z-axis. Tracks spiral in the x-y plane (radius $\alpha p_T / B$) and advance linearly in z, with random starting angles and radii to mimic varying momenta. This concept is fundamental in HEP for tracking studies, where curvature yields momentum and vertex positions aid reconstruction, such as in heavy-flavor decays. Visualizing tracks within the geometry tests detector coverage, inspiring event-display figures in research papers. Using OpenGL (Draw("ogl")) for interactive 3D rendering, with outline style and camera rotation, provides a clear view of detector components and tracks. Transparency and color settings enhance visibility. This visualization technique is vital for illustrating detector design or simulation validation, often used in technical notes or conference talks to convey experimental setup.



This project can inspire a "Detector Simulation" section with a 3D figure showing STAR's layout or track reconstruction. It could be extended with real GEANT geometry or integrated with event generators to show realistic events. The figure is ideal for discussing acceptance or tracking efficiency, and the code's modularity allows adding more detector layers (such as calorimeters) for a more complete model, enhancing its academic value.

II. INTRODUCTION TO PYTHIA8

Pythia8 is mainly associated with event generation from first principles. Basically, it is a Monte Carlo event generator that simulates high-energy collisions between fundamental particles, such as those occurring at the Large Hadron Collider or the Relativistic Heavy Ion Collider. Pythia8 enables theorists and experimentalists to generate "truth-level" events, which serve as reference for detector simulations and the main validation tool used to validate or refute physics models. Its modular C++ structure, customizability, and compatibility with detector-level frameworks make it a foundational tool in both STAR and CERN research environments.

Proton-proton (pp) collisions occur when two protons are accelerated to high energies and smashed together in particle accelerators like the Large Hadron Collider at CERN or the Relativistic Heavy Ion Collider at Brookhaven. At everyday energies, protons behave as stable composite particles made of three valence quarks (two up and one down) bound by gluons, the carriers of the strong nuclear force described by Quantum Chromodynamics. However, in high-energy collisions (such as $\sqrt{s} = 13$ TeV at LHC or 200 GeV at RHIC), the protons' internal structure becomes relevant: they are clouds of quarks, antiquarks, and gluons (collectively called partons) whose distributions are encoded in Parton Distribution Functions. Collisions can be classified as elastic, where protons scatter intact, or inelastic, where protons break apart (producing new particles), or diffractive, where one or both protons remain intact but exchange momentum via Pomeron exchange, a QCD phenomenon. Inelastic collisions dominate at high energies, leading to a cascade of processes from initial parton interactions to final hadron production.

The total cross-section for pp collisions at LHC energies is around 100 millibarns, with QCD processes contributing the majority due to the strong force's dominance over electroweak interactions. Key observables include particle multiplicity (number of produced particles), transverse momentum (p_T) spectra, and correlations, which probe the proton's structure and QCD dynamics. These collisions serve as a baseline for studying heavy-ion physics, such as the quark-gluon plasma, and searching for new physics, but QCD effects like multiple parton interactions can create backgrounds that mimic rare signals. In contrast to soft QCD, hard processes involve high p_T (> few GeV/c) where α_s is small (0.1–0.3), allowing pQCD calculations. These include jet production from 2→2 parton scatters (e.g., gg→gg, qg→qg). Parton showers add perturbative corrections, modeled as branching processes. Hard QCD probes α_s evolution and gluon PDFs, dominant at low x, where x is parton momentum fraction. At LHC, hard processes like inclusive jets or multijets are measured differentially, constraining PDFs and extracting α_s at Z-boson mass scale with NNLO accuracy. Factorization separates hard matrix elements from PDFs and showers, enabling predictions. Hard QCD constitutes backgrounds for new physics but also tests QCD at high scales, with some correlations being sensitive to new colored particles. Beyond QCD, electroweak processes occur, though with smaller cross-sections. These include Drell-Yan processes, W/Z production, and Higgs boson decays. QCD contributes to these via higher-order corrections (such as gluon radiation) and backgrounds (such as multijet fakes).

In pp collisions, new physics searches (e.g., supersymmetry, extra dimensions) rely on QCD modeling for backgrounds. Collective phenomena in high-multiplicity pp events suggest QGP-like effects, blurring lines with heavy-ion collisions, possibly from hydrodynamic flow or string shoving. Recent advances in QCD theory, such as reconciling quantum and relativistic aspects, improve proton structure understanding, linking collisions to internal dynamics via better data interpretation. Overall, pp collisions are a fundamental in QCD, with soft processes dominating bulk production and hard ones probing high scales, essential for precision physics at future colliders.

A. Proton-Proton collisions

This project on proton-proton collisions code uses Pythia8, a Monte Carlo event generator, to simulate proton-proton (pp) collisions at 200 GeV, mimicking conditions at the Relativistic Heavy Ion Collider (RHIC) for the STAR experiment. A critical concept is the simulation of hard QCD processes (enabled via HardQCD:all = on), which involve high transverse momentum (p_T) parton scatterings (e.g., quark-gluon interactions) described by perturbative QCD (pQCD). Pythia8 models the full event evolution, including initial-state radiation, parton showers, multiple parton interactions, and hadronization via the Lund string model, transforming partons into observable hadrons. This allows researchers to predict particle spectra and test theoretical models against experimental data, a cornerstone of HEP simulation workflows.



```

abdalrah_waliid@DESKTOP-JI ~ + ~
abdalrah_waliid@DESKTOP-JI:~/Pythia8315$ ./ppcollision
*----- PYTHIA Process Initialization -----*
| We collide p+ with p+ at a CM energy of 2.000e+02 GeV |
|----- PYTHIA Subprocesses -----|
|----- PYTHIA Multiparton Interactions Initialization -----|
|----- PYTHIA Flag + Mode + Parm + Word + FVec + MVec + PVec + WVec Settings (changes only) -----|
|----- PYTHIA Particle Data Table (changed only) -----|
no particle data has been changed from its default value

```

```

abdallah_walid@DESKTOP-JI ~ + ~
no particle data has been changed from its default value
----- End PYTHIA Particle Data Table -----
----- PYTHIA Info Listing -----
Beam A: id = 2212, px = 1.000e+02, e = 1.000e+02, m = 9.383e-01.
Beam B: id = 2212, px = -1.000e+02, e = 1.000e+02, m = 9.383e-01.
In 1: id = 21, x = 4.163e-02, pdf = 1.475e+00 at Q2 = 1.075e+00.
In 2: id = 1, x = 4.311e-02, pdf = 3.665e-01 at same Q2.

Subprocess q g -> q g with code 113 is 2 -> 2.
It has sHat = 7.177e+01, tHat = -1.091e+00, uHat = -7.066e+01,
    pTHat = 1.035e+00, mTHat = 0.000e+00, mDTHat = 1.037e+00,
    thetaTHat = 2.473e-01, phiTHat = 2.019e+00,
    alphaTHat = 7.448e-03, alphaH = 4.931e-01 at Q2 = 1.075e+00.

Impact parameter b = 8.727e-01 gives enhancement factor = 1.098e+00.
Max pt scale for MPI = 1.037e+00, ISR = 1.037e+00, FSR = 1.037e+00.
Number of MPI = 2, ISR = 1, FSProc = 2, FSReson = 0.

----- End PYTHIA Info Listing -----
----- PYTHIA Event Listing (hard process) -----
no id name status mothers daughters colours p_x p_y p_z e m
0 90 (system) -11 0 0 0 0 0 0.000 0.000 0.000 200.000 200.000
1 2212 (p+) -12 0 0 3 0 0 0.000 0.000 99.996 100.000 0.938
2 2212 (p+) -12 0 0 22 0 0 0.000 0.000 99.996 100.000 0.938
3 21 (g) -21 1 0 5 6 102 103 0.000 0.000 -99.996 100.000 0.938
4 1 (d) -21 2 0 5 6 101 0 0.000 0.000 -11.311 4.311 8.000
5 21 g 23 3 4 0 101 103 -0.499 0.933 4.028 4.158 0.000
6 1 d 23 3 4 0 102 0 0.049 -0.933 -1.176 4.315 0.330
Charge sum: -0.333 Momentum sum: 0.000 0.000 -0.148 8.473 8.472

----- End PYTHIA Event Listing -----
----- PYTHIA Event Listing (complete event) -----
no id name status mothers daughters colours p_x p_y p_z e m
0 90 (system) -11 0 0 0 0 0 0.000 0.000 0.000 200.000 200.000
1 2212 (p+) -12 0 0 22 0 0 0 0.000 0.000 99.996 100.000 0.938
2 2212 (p+) -12 0 0 23 0 0 0 0.000 0.000 99.996 100.000 0.938
3 21 (g) -21 7 0 5 6 102 103 0.000 0.000 4.163 4.163 0.000
4 1 (d) -21 8 0 5 6 101 0 0.000 0.000 -1.311 4.311 8.000
5 21 (g) -23 3 4 9 101 103 0 -0.933 4.028 4.158 0.000
6 1 (d) -23 3 4 10 102 0 0.049 -0.933 -1.176 4.311 8.000
7 2 (u) -41 22 22 11 3 102 0 -0.000 -0.000 12.369 12.369 0.000
8 1 (d) -42 17 17 4 4 101 0 0.000 0.000 -4.311 4.311 0.000
9 21 (g) -44 5 5 12 13 101 103 0.018 1.467 4.011 4.271 0.000
10 1 (d) -44 6 6 25 25 102 0 0.457 -0.924 -4.091 4.232 0.330
11 2 (u) -43 7 0 14 14 103 0 -0.075 0.545 8.156 8.177 0.330
12 21 (g) -51 9 9 15 15 101 103 0 -0.208 6.613 6.601 0.000
13 21 (g) -51 9 0 27 27 101 103 0.105 0.293 4.181 0.368 0.000
14 2 (u) -52 11 11 26 26 103 0 -0.345 -0.378 5.956 5.987 0.330
15 21 (g) -51 12 0 24 24 105 104 -0.688 0.801 4.792 4.998 0.000
16 21 (g) -51 12 0 28 28 101 105 0.071 0.207 1.153 1.262 0.000
17 1 (d) -53 23 23 8 8 101 0 0.000 0.000 -4.379 4.379 0.000
18 21 (g) -53 29 29 26 21 106 107 0.606 0.000 0.336 0.363 0.000
19 21 (g) -53 30 30 29 21 106 108 0.000 0.000 -0.050 0.050 0.000
20 21 (g) -53 18 18 19 31 106 109 0.013 0.309 0.218 0.379 0.000
21 21 (g) -53 18 19 32 32 109 108 -0.013 0.309 -4.965 4.974 0.000
22 2 (u) -61 1 0 7 7 102 0 -0.851 0.762 12.386 12.433 0.000
23 1 (d) -61 2 0 17 17 109 0 -1.249 -0.302 -4.305 4.493 0.000
24 21 (g) -62 15 15 53 53 105 104 -1.046 1.098 4.826 5.053 0.000
25 1 (d) -62 10 10 41 41 102 0 -0.729 -1.206 -4.616 4.263 0.330
26 2 (u) -62 10 10 51 51 103 0 -0.703 -0.923 6.032 6.030 0.330
27 21 (g) -62 13 13 52 52 104 103 0.062 0.303 0.152 0.240 0.000
28 21 (g) -62 16 16 54 54 101 105 0.376 0.275 1.099 1.186 0.000
29 21 (g) -61 1 0 18 18 108 108 0.450 0.513 -0.046 0.683 0.000
30 21 (g) -61 2 0 19 19 107 107 0.345 0.004 -5.037 5.049 0.000
31 21 (g) -62 20 20 55 55 106 101 0.447 0.195 0.124 0.504 0.000
32 21 (g) -62 21 21 56 56 109 105 0.348 0.322 -5.030 5.060 0.000
33 2203 (uu_1) -63 1 0 42 42 104 102 0 -0.170 55.555 55.563 0.777
34 1 (d) -63 1 0 37 40 111 0 -0.272 -0.302 29.645 29.658 0.230
35 -2 (ubar) -63 1 0 37 40 111 0 0.111 0.207 -0.820 1.859 2.069 0.330
36 2203 (uu_1) -63 2 0 57 57 109 0 0.904 0.297 -90.052 90.068 0.771
37 111 (pi0) -83 34 35 88 89 0 0 -0.035 -0.670 11.471 11.491 0.135
38 -211 pi- 83 34 35 0 0 0 0 -0.195 0.021 14.037 14.039 0.140
39 223 (omega) -84 34 35 99 92 0 0 -0.187 0.049 2.982 3.099 0.822
40 111 (pi0) -84 34 35 93 94 101 0 0.337 -0.562 5.973 3.090 0.135
41 111 (pi0) -71 35 35 19 59 102 0 -0.729 -1.056 -4.518 4.523 0.770
42 2203 (uu_1) -71 33 33 43 59 102 0 0.467 -0.112 55.555 55.563 0.771
43 113 (rho0) -83 41 42 70 71 0 0 -0.244 -0.242 -1.603 1.707 0.474
44 111 (pi0) -83 41 42 95 96 0 0 -0.015 0.019 -0.180 0.227 0.135
45 313 (K0) -83 41 42 72 73 0 0 -0.527 -0.951 -1.361 2.096 1.165

```

```

abdallah_walid@DESKTOP-JI ~ + abdallah_walid@DESKTOP-JI ~ + abdallah_walid@DESKTOP-JI ~ +
45   313 (K*)          -83  41  42  72  73  0  0  -0.527  -0.951  -1.361  2.096  1.165
46   -323 (K*-)        -83  41  42  74  75  0  0  -0.036  0.128  0.216  0.917  0.681
47   2212 p+           84  41  42  0  0  0  0  -0.252  -0.017  1.662  1.925  0.938
48   -2212 p-           84  41  42  0  0  0  0  0.046  0.097  7.695  7.730  0.938
49   211 pi+           84  41  42  0  0  0  0  0.119  -0.518  0.322  9.838  1.140
50   2214 (Delta+)     -84  41  42  76  77  0  0  -0.075  0.176  35.898  35.923  1.331
51   2 (u)             -71  26  26  58  69  103  0  0  -0.763  -0.068  6.023  6.088  0.330
52   21 (g)            -71  27  27  58  69  104  103  0.062  0.303  0.152  0.344  0.000
53   21 (g)            -71  24  24  58  69  105  104  -1.016  1.098  4.820  5.053  0.000
54   21 (g)            -71  28  28  58  69  101  105  0.376  0.275  1.099  1.186  0.000
55   21 (g)            -71  31  31  58  69  106  105  0.047  0.047  0.105  0.100  0.000
56   21 (g)            -71  32  32  58  69  105  105  0.303  0.322  0.307  0.299
57   2203 (uu_1)       -71  36  36  58  69  109  0.000  0.984  0.297  -90.052  98.066  0.771
58   211 pi+           83  51  57  0  0  0  0  -0.392  0.689  2.553  2.677  0.140
59   -213 (rho-)        83  51  57  78  79  0  0  -0.569  -0.038  3.798  3.910  0.774
60   323 (K**)          83  51  57  80  81  0  0  0.019  0.118  1.706  1.935  0.906
61   -311 (Kbar0)      83  51  57  82  82  0  0  -0.557  0.525  2.328  2.561  0.498
62   -211 pi-           83  50  57  0  0  0  0  0.005  0.560  0.468  1.179  0.140
63   323 (rho+)         83  51  57  53  84  0  0  0.209  0.209  -0.994  0.940
64   -211 pi-           84  51  57  0  0  0  0  0.178  0.563  0.564  0.829  0.140
65   111 (pi0)          84  51  57  97  98  0  0  0.084  -0.093  -4.852  4.056  0.135
66   211 pi+           84  51  57  0  0  0  0  0.049  -0.089  -16.216  16.217  0.140
67   223 (omega)        84  51  57  99  101  0  0  0.291  -0.153  -11.269  11.291  0.769
68   2214 (Delta+)     -84  51  57  85  86  0  0  0.281  0.194  -31.143  31.169
69   331 (eta*)         -80  51  162  168  0  0  0.000  0.000  -3.075  31.078  0.335
70   2211 pi+           91  43  0  0  0  0  0  -0.044  -0.169  -1.259  1.278  0.140
71   -211 pi-           91  43  0  0  0  0  0  -0.208  -0.073  -0.300  0.428  0.140
72   321 K+             91  45  0  0  0  0  0  -0.434  -0.982  -0.619  1.334  0.494
73   -211 pi-           91  45  0  0  0  0  0  -0.093  0.031  -0.743  0.762  0.140
74   -311 (Kbar0)      91  46  0  87  87  0  0  0.294  0.144  0.126  0.669  0.498
75   -211 pi-           91  46  0  0  0  0  0  0.259  -0.016  0.099  0.388  0.140
76   2212 p+           91  58  0  0  0  0  0  -0.119  0.119  19.771  19.976  0.378
77   -211 (pi0)         91  59  0  105  106  0  0  0.085  0.004  16.327  16.327  0.135
78   -211 pi-           91  59  0  0  0  0  0  -0.327  0.182  0.688  0.795  0.140
79   111 (pi0)          91  59  0  107  108  0  0  -0.241  -0.140  3.189  3.265  0.135
80   321 K+             91  60  0  0  0  0  0  -0.165  0.318  1.692  1.251  0.494
81   111 (pi0)          91  60  0  109  110  0  0  0.184  -0.199  0.614  0.684  0.135
82   310 (K_S0)         91  61  61  111  112  0  0  -0.557  0.525  2.328  2.561  0.498
83   211 pi+            91  63  0  0  0  0  0  0.071  0.112  0.189  0.112  0.140
84   111 (pi0)          91  63  0  111  111  0  0  0.119  -0.331  0.765  0.642  0.135
85   2212 p+           91  68  0  0  0  0  0  0.178  0.030  -18.956  18.988  0.938
86   111 (pi0)          91  68  0  115  116  0  0  0.103  0.163  -12.187  12.189  0.135
87   310 (K_S0)         91  74  74  117  118  0  0  -0.294  0.144  0.126  0.669  0.498
88   22 gamma           91  37  0  0  0  0  0  0.021  -0.103  1.193  1.197  0.000
89   22 gamma           91  37  0  0  0  0  0  0.056  -0.567  10.278  10.293  0.000
90   211 pi+            91  39  0  0  0  0  0  0.012  0.012  0.003  0.148  0.140
91   -211 pi-           91  39  0  0  0  0  0  0.024  -0.207  0.079  2.097  0.140
92   111 (pi0)          91  39  0  119  120  0  0  0.066  0.093  0.363  0.483  0.135
93   22 gamma           91  40  0  0  0  0  0  0.184  -0.184  1.386  1.326  0.000

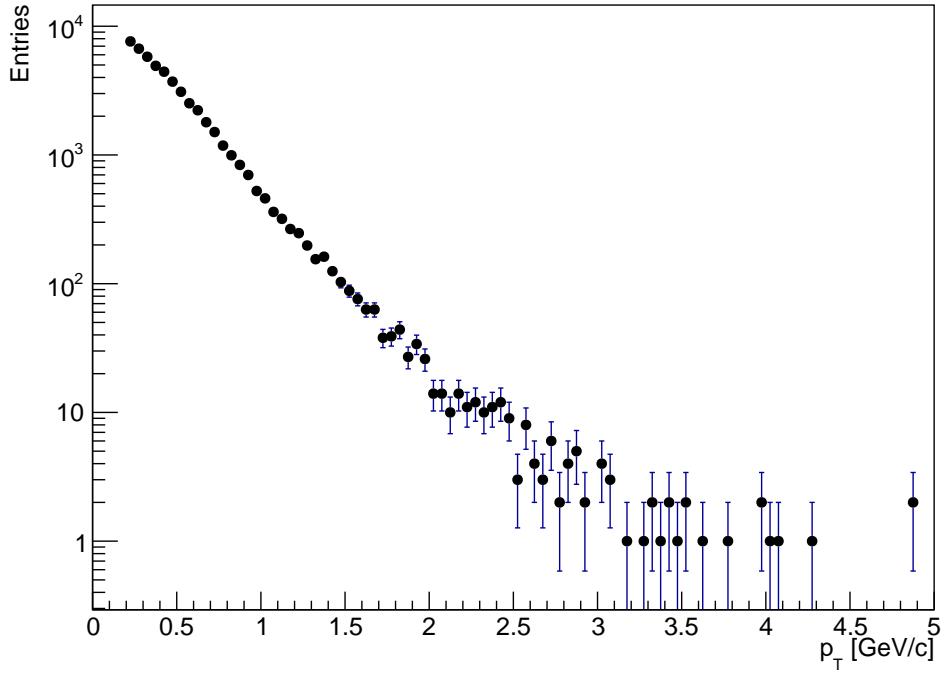
----- End PYTHIA Event Listing -----
Pythia::next(): 1000 events have been generated
Pythia::next(): 2000 events have been generated
Pythia::next(): 3000 events have been generated
Pythia::next(): 4000 events have been generated
Info in <TCanvas::Print>: pdf file pt_distribution.pdf has been created
Info in <TCanvas::Print>: png file pt_distribution.png has been created
Info in <TCanvas::Print>: pdf file eta_distribution.pdf has been created

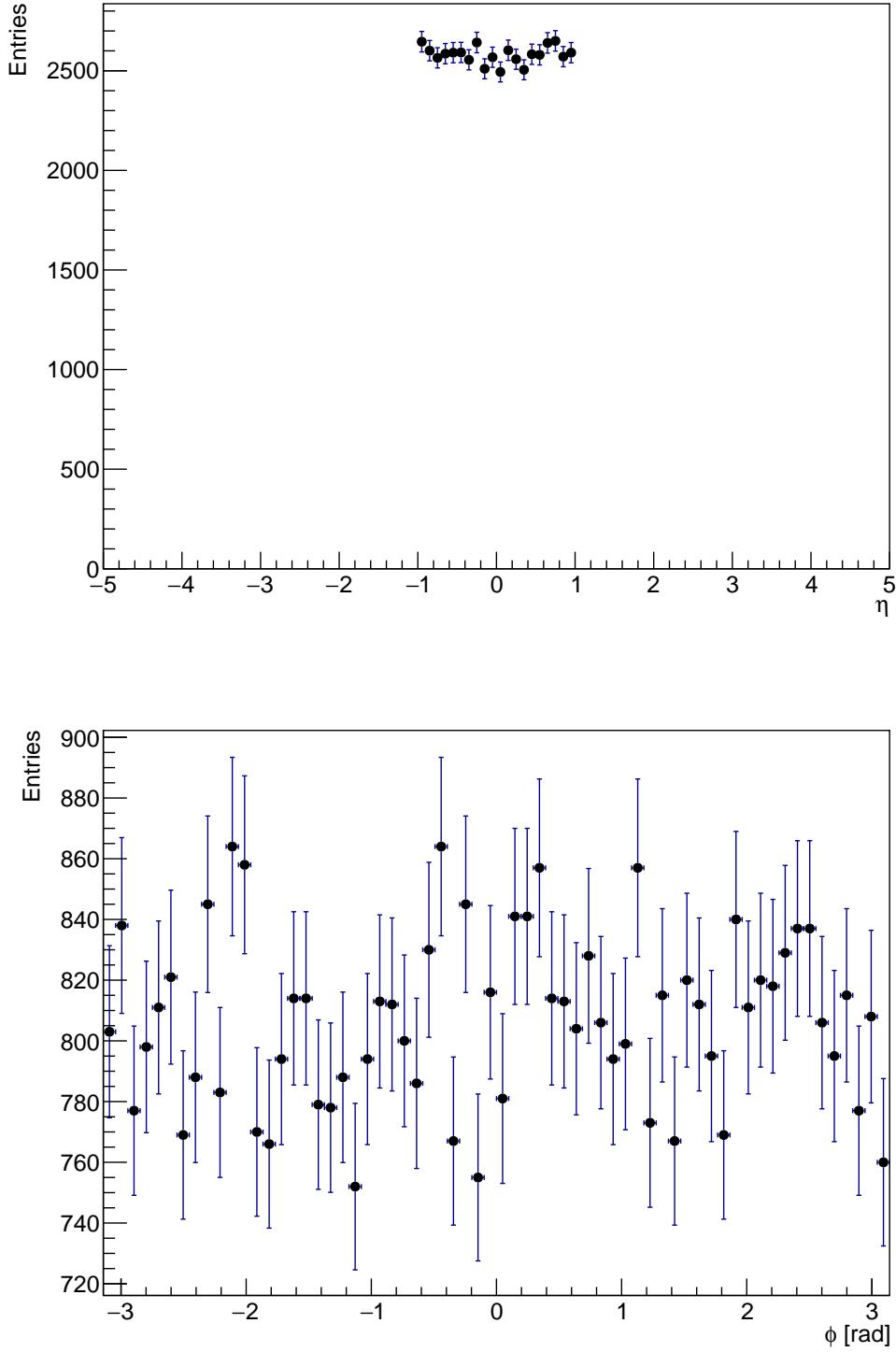
```

```

abdallah_walid@DESKTOP-JI ~ + ~
Pythia::next(): 1000 events have been generated
Pythia::next(): 2000 events have been generated
Pythia::next(): 3000 events have been generated
Pythia::next(): 4000 events have been generated
Info in <TCanvas::Print>: pdf file pt_distribution.pdf has been created
Info in <TCanvas::Print>: png file pt_distribution.png has been created
Info in <TCanvas::Print>: pdf file eta_distribution.pdf has been created
Info in <TCanvas::Print>: png file eta_distribution.png has been created
Info in <TCanvas::Print>: pdf file phi_distribution.pdf has been created
Info in <TCanvas::Print>: png file phi_distribution.png has been created
Info in <TCanvas::Print>: pdf file pt_vs_eta.pdf has been created
Info in <TCanvas::Print>: png file pt_vs_eta.png has been created
----- PYTHIA Event and Cross Section Statistics -----
| Subprocess | Code | Number of events | sigma +- delta | (estimated) (mb) |
|            |     | Tried Selected Accepted |           |           | |
|---|---|---|---|---|---|
| g g -> g g | 111 | 12388 | 24112 | 21112 | 3.507e+01 3.770e-01 |
| g g -> q qbar (uds) | 112 | 180 | 25 | 25 | 3.177e-01 3.266e-01 |
| g g -> q' q' | 113 | 11113 | 2143 | 2148 | 2.887e+01 3.798e-01 |
| q qbar -> q q (bar)' | 114 | 2514 | 413 | 413 | 5.509e+00 1.412e-01 |
| q qbar -> g g | 115 | 9 | 1 | 1 | 3.817e-02 3.817e-02 |
| q qbar -> q' qbar' (uds) | 116 | 3 | 1 | 1 | 1.409e-02 1.409e-02 |
| g g -> c cbar | 121 | 31 | 5 | 5 | 7.538e-02 2.247e-02 |
| q qbar -> c cbar | 122 | 1 | 0 | 0 | 0.000e+00 0.000e+00 |
| g g -> b bbar | 123 | 2 | 0 | 0 | 0.000e+00 0.000e+00 |
| q qbar -> b bbar | 124 | 0 | 0 | 0 | 0.000e+00 0.000e+00 |
| sum | | 26108 | 5000 | 5000 | 6.689e+01 5.465e-01 |
----- End PYTHIA Event and Cross Section Statistics -----
----- PYTHIA Error and Warning Messages Statistics -----
| times message |
| 0 no errors or warnings to report |
----- End PYTHIA Error and Warning Messages Statistics -----
abdallah_walid@DESKTOP-JI ~ + ~

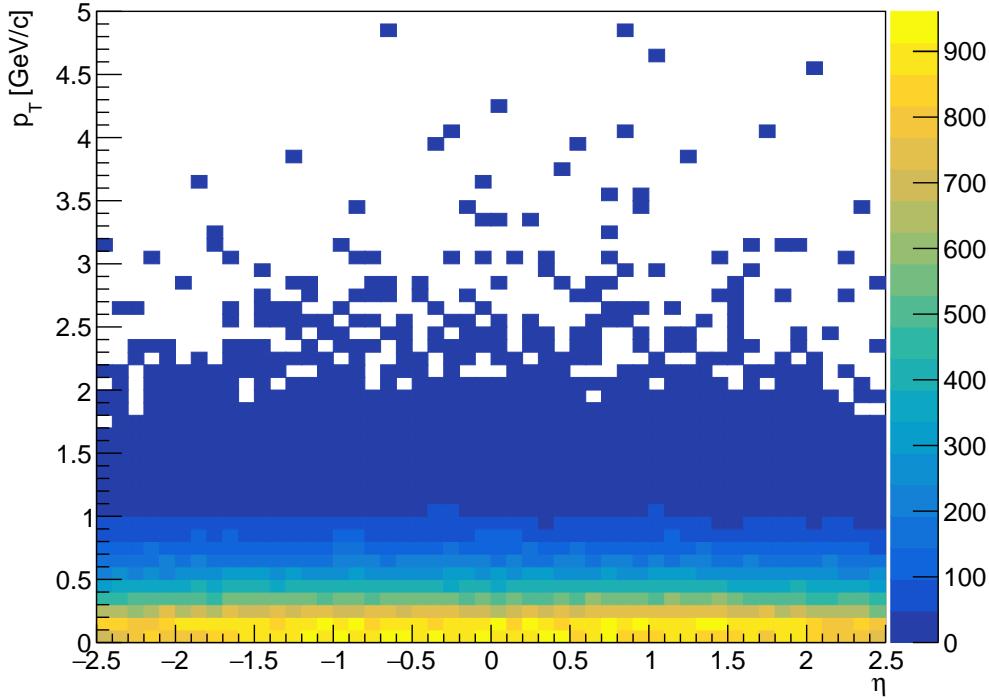
```





The code generates histograms for key kinematic variables— p_T (transverse momentum), η (pseudorapidity), and ϕ (azimuthal angle)—using ROOT's histogram classes (TH1F, TH2F). A crucial concept is the application of detector acceptance cuts ($|\eta| < 1$, $p_T > 0.2$ GeV/c), reflecting STAR's central tracking capabilities. These cuts filter particles to those detectable by the experiment, enabling realistic comparisons with data. The histograms

capture distributions: p_T follows a steep exponential (soft) to power-law (hard) shape, η is flat in the central rapidity plateau, and ϕ is uniform due to cylindrical symmetry. The 2D η vs p_T histogram below shows full kinematic coverage, highlighting detector limitations, which is vital for understanding experimental biases in papers.



ROOT's visualization tools can produce publication-quality plots with logarithmic axes for p_T (to emphasize tails) and a color map for 2D plots. The use of `Sumw2()` ensures proper statistical error propagation (Poisson uncertainties), critical for HEP analyses where precision matters. The code's output, including Pythia statistics, provides insights into process contributions (e.g., $gg \rightarrow gg$ dominance). This workflow—simulate, filter, histogram, visualize—is standard for validating Monte Carlo predictions against data, inspiring sections in papers on simulation methods or kinematic results.

This project exemplifies how to simulate collider events, apply experimental constraints, and analyze distributions. It could inspire a methodology section detailing Monte Carlo tools, a results section with kinematic plots, or comparisons with STAR data (e.g., p_T spectra fitted with Tsallis functions). Extending the code by adding particle identification or soft QCD processes could enhance its applicability to specific physics questions, such as jet production or bulk properties in pp collisions.

III. REFERENCES

- ¹ROOT Team, *ROOT Data Analysis Framework: Users Guide* (2024), <https://root.cern/manual/>.
- ²ROOT Team, *ROOT Class Reference* (2024), <https://root.cern/doc/master/>.
- ³G. Cowan, *Statistical Data Analysis* (Oxford University Press, 1998).
- ⁴L. Lyons, *Statistics for Nuclear and Particle Physicists* (Cambridge University Press, 1986).
- ⁵Particle Data Group, “Review of particle physics,” *Prog. Theor. Exp. Phys.* **2024**, 083C01 (2024), <https://pdg.lbl.gov/>.
- ⁶R. Ellis, W. Stirling, and B. Webber, “Qcd and collider physics,” in *Cambridge Monographs on Particle Physics, Nuclear Physics and Cosmology* (Cambridge University Press, 1996).
- ⁷CMS Collaboration, “Top quark physics at the lhc,” arXiv preprint (2014), arXiv:1401.6075.
- ⁸C. Heinz, *The Listings Package* (2023), <https://ctan.org/pkg/listings>.
- ⁹U. Kern, *The xcolor package* (2023), <https://ctan.org/pkg/xcolor>.
- ¹⁰J. Ellis, *tikz-feynman: Feynman diagrams with TikZ* (2023), <https://ctan.org/pkg/tikz-feynman>.
- ¹¹Wikibooks contributors, “Latex wikibook,” (2024), <https://en.wikibooks.org/wiki/TeX>.
- ¹²B. Frieden, *Probability, Statistical Optics, and Data Testing* (Springer, 2001).
- ¹³CMS Collaboration, “Measurement of the cms beam spot during lhc run 2,” *Tech. Rep. CMS-DP-2016-070* (CERN, 2016).
- ¹⁴C. Collaboration, “Measurement of the underlying event activity in proton-proton collisions at 13 tev with the cms detector,” *Journal of High Energy Physics* **2015**, 008 (2015), available at <https://arxiv.org/abs/1512.00815>, arXiv:1512.00815.
- ¹⁵T. Sjöstrand, S. Mrenna, and P. Skands, “A brief introduction to pythia 8.1,” *Computer Physics Communications* **178**, 852–867 (2008).
- ¹⁶A. Collaboration, “Enhanced production of multi-strange hadrons in high-multiplicity proton-proton collisions,” *Nature Physics* **13**, 535–539 (2017).
- ¹⁷H1 and Z. Collaborations, “Combination of hera deep inelastic scattering data and impact on pdfs,” *European Physical Journal C* **80**, 118 (2020).
- ¹⁸S. Collaboration, “Charged particle distributions in pp collisions at $\sqrt{s} = 200$ gev,” *Physical Review D* **74**, 032006 (2005).
- ¹⁹R. Brun and F. Rademakers, “Root — an object-oriented data analysis framework,” *Nuclear Instruments and Methods in Physics Research Section A* **389**, 81–86 (1997).
- ²⁰P. Skands *et al.*, “Qcd for collider physics,” *Annual Review of Nuclear and Particle Science* **64**, 159–186 (2014).
- ²¹R. Barlow, *Statistics: A Guide to the Use of Statistical Methods in the Physical Sciences* (John Wiley & Sons, 2002).
- ²²C. Collaboration, “Measurement of charged particle spectra in pp collisions at $\sqrt{s} = 13$ tev,” *Journal of High Energy Physics* **2018**, 130 (2018).
- ²³S. Collaboration, “The star time projection chamber: A unique tool for studying high multiplicity events at rhic,” *Nuclear Instruments and Methods in Physics Research Section A* **499**, 659–678 (2003).
- ²⁴S. Agostinelli *et al.*, “Geant4—a simulation toolkit,” *Nuclear Instruments and Methods in Physics Research Section A* **506**, 250–303 (2003).
- ²⁵R. Brun *et al.*, “Root geometry package,” in *Proceedings of CHEP 2003* (2003) available via CERN Document Server or ROOT documentation: <https://root.cern/doc/master/>.
- ²⁶S. Collaboration, “ ρ^0 production and possible modification in au+au and pp collisions at $\sqrt{s_{NN}} = 200$ gev,” *Physical Review Letters* **99**, 112301 (2007).
- ²⁷M. E. Peskin and D. V. Schroeder, *An Introduction to Quantum Field Theory* (Addison-Wesley, 1995).

Appendix A: Codes

This appendix shows all the codes used for each project, with comments added to improve the reader's understanding of the logic and flow of the codes.

1. Importing randomly generated data into a simple histogram

Listing 1. extractandplot.c

```

1 void extractandplot()
2 {
3
4     //Create histogram and random number generator
5     TH1F *hist = new TH1F("hist", "", 100, 0, 15);      // Histogram
6     with 100 bins, range 0 15
7     TRandom2 *rand = new TRandom2(3);                  // Random
8     generator with seed=3
9     fstream file;
10
11    // Generate random Gaussian-distributed data and save to file
12    file.open("data.txt", ios::out);
13    for (int i = 0; i < 1000; i++)
14    {
15        double r = rand->Gaus(5, 1);      // Gaussian with mean=5, sigma
16        =1
17        file << r << endl;                // Write values to file
18    }
19    file.close();
20
21
22    // Read data back from file and fill histogram
23    file.open("data.txt", ios::in);
24    double value;
25    while (1)
26    {
27        file >> value;
28        hist->Fill(value);              // Fill histogram with each
29        value
30        if (file.eof()) break;          // Stop at end of file
31    }
32    file.close();
33
34    // Customize histogram appearance
35    hist->SetFillColor(kGreen - 9);
36    hist->GetXaxis()->SetTitle("Distribution");
37    hist->GetYaxis()->SetTitle("Entries");
38    hist->GetXaxis()->SetTitleSize(0.05);
39    hist->GetYaxis()->SetTitleSize(0.05);
40    hist->GetXaxis()->SetLabelSize(0.05);
41    hist->GetYaxis()->SetLabelSize(0.05);
42
43
44    // Define Gaussian fit function with initial parameters
45    TF1 *fit = new TF1("fit", "gaus", 0, 15);
46    fit->SetLineWidth(3);
47    fit->SetLineColor(kRed);
48    fit->SetLineStyle(2);
49    fit->SetParameter(0, 40);      // Normalization (initial guess)
50    fit->SetParameter(1, 5);       // Mean (initial guess)

```

```

47     fit->SetParameter(2, 1);      // Sigma (initial guess)
48
49
50 // Create canvas, enable grid, fit histogram with Gaussian, and
51 // draw
52 TCanvas *c1 = new TCanvas();
53 c1->SetTickx();
54 c1->SetTicky();
55 c1->SetGridx();
56 c1->SetGridy();
57 hist->Fit("fit");    // Perform Gaussian fit
58 hist->Draw();        // Draw histogram
59
60 // Add legend to distinguish data and fit
61 TLegend *leg = new TLegend(0.5, 0.55, 0.8, 0.75);
62 leg->SetBorderSize(0);
63 leg->AddEntry(hist, "Measured Data", "p");
64 leg->AddEntry(fit, "Fit Function", "l");
65 leg->Draw();
66
67 // Draw reference line and annotation arrow
68 TLine *l = new TLine(0, 20, 15, 20);    // Horizontal line at y=20
69 l->SetLineWidth(2);
70 l->SetLineColor(kOrange);
71 l->Draw();
72
73 //Arrow pointing to a specific histogram bin
74 double x0 = 6.3;
75 int bin = hist->FindBin(x0);
76 double y0 = hist->GetBinContent(bin);
77 TArrow *arr = new TArrow(10, 30, x0, y0);
78 arr->SetLineWidth(2);
79 arr->SetArrowSize(0.02);
80 arr->SetLineColor(kBlue);
81 arr->Draw();
82
83 //Text label near the arrow
84 TLatex *t = new TLatex(10, 30, "Important");
85 t->Draw();
86
87 //Extract fit parameters and compute mean/sigma ratio
88 double mean = fit->GetParameter(1);
89 double sigma = fit->GetParameter(2);
90 cout << mean / sigma << endl;    // Print ratio to console
91 }
```

2. Analyzing the J/ψ resonance

Listing 2. extract2.c

```

1 void extract2(const char* infile = "simulated_mass.txt") {
2     // --- Settings ---
3     const int nBins = 100;
4     const double minMass = 2.0;      // GeV/c^2
5     const double maxMass = 4.0;      // GeV/c^2
6
7     // --- Histogram ---
8     TH1F *hMass = new TH1F("hMass", ";Invariant_Mass_{#mu#mu}(GeV/
9         c^2);Entries/bin", nBins, minMass, maxMass);
10    hMass->SetMarkerStyle(20);
11    hMass->SetMarkerSize(1.0);
12
13    // --- Random data simulation (replace later with real data
14        reading) ---
15    TRandom3 rand(42);
16    std::ofstream fout(infile);
17    const double signalYield = 500;
18    const double bkgYield = 2000;
19
20    // Signal: Gaussian centered at 3.1 GeV/c^2 (J/psi mass)
21    for (int i = 0; i < signalYield; i++) {
22        double mass = rand.Gaus(3.097, 0.05); // mean, sigma
23        fout << mass << "\n";
24    }
25    // Background: exponential falloff
26    for (int i = 0; i < bkgYield; i++) {
27        double mass = minMass - log(rand.Uniform()) * 0.8; // lambda
28        if (mass < maxMass) fout << mass << "\n";
29    }
30    fout.close();
31
32    // --- Read data ---
33    std::ifstream fin(infile);
34    double val;
35    while (fin >> val) {
36        if (val >= minMass && val <= maxMass) hMass->Fill(val);
37    }
38    fin.close();
39
40    // --- Fit function: Gaussian + exponential background ---
41    TF1 *fitFunc = new TF1("fitFunc", "[0]*exp([1]*x)+[2]*exp(
42        (-0.5*((x-[3])/[4])*2)", minMass, maxMass);
43    fitFunc->SetParameters(100, -1, 200, 3.1, 0.05); // initial
44        guesses
45    fitFunc->SetParNames("BkgNorm", "BkgSlope", "SigNorm", "SigMean",
46        "SigSigma");
47
48    // --- Canvas ---
49    TCanvas *c1 = new TCanvas("c1", "STAR-style Analysis", 800, 600);
50    gStyle->SetOptFit(1111);
51    gStyle->SetStatX(0.88);
52    gStyle->SetStatY(0.88);
53
54    hMass->Draw("E1");
55    hMass->Fit("fitFunc", "R");
56
```

```
52 // --- Legend ---
53 TLegend *leg = new TLegend(0.2, 0.55, 0.48, 0.75);
54 leg->SetBorderSize(0);
55 leg->SetFillStyle(0);
56 leg->AddEntry(hMass, "Simulated Data", "lep");
57 leg->AddEntry(fitFunc, "Fit: Gauss+Exp", "l");
58 leg->Draw();
59
60 // --- Annotation ---
61 TLatex text;
62 text.SetNDC();
63 text.SetTextSize(0.04);
64 text.DrawLatex(0.15, 0.85, "#bf{STAR Simulation}");
65 text.DrawLatex(0.15, 0.80, "p+p #sqrt{s} = 200 GeV");
66
67 // --- Save outputs ---
68 c1->SaveAs("mass_spectrum.pdf");
69 c1->SaveAs("mass_spectrum.png");
70
71 // --- Save histogram and fit to ROOT file ---
72 TFile outFile("analysis_results.root", "RECREATE");
73 hMass->Write();
74 fitFunc->Write();
75 outFile.Close();
76 }
```

3. Importing and Analyzing Data from multiple files

Listing 3. multiplefilesupgrade.c

```

1 #include <TRandom3.h>
2 #include <TStyle.h>
3 #include <TLegend.h>
4 #include <THStack.h>
5 #include <TH1F.h>
6 #include <TF1.h>
7 #include <TCanvas.h>
8 #include <TFile.h>
9 #include <iostream>
10 #include <fstream>
11
12 void multiplefilesupgrade() {
13     gStyle->SetOptStat(0);
14
15     const int nHists = 6;           // number of histograms per file
16     const int nFiles = 10;          // total number of files
17     const int nEventsPerFile = 1000; // events per file
18
19     TRandom3 randGen(0); // random seed for reproducibility
20
21     // Parameters for each dataset's Gaussian (mean, sigma)
22     double means[nHists] = {0.0, 0.5, -0.5, 1.0, -1.0, 0.0};
23     double sigmas[nHists] = {1.0, 0.8, 1.2, 0.6, 1.5, 2.0};
24
25     TH1F *hist[nHists];
26     THStack *hstack = new THStack("stack", "Momentum Distributions;P_{T} [GeV/c];Entries");
27     TLegend *leg = new TLegend(0.65, 0.5, 0.88, 0.88);
28     leg->SetBorderSize(0);
29     leg->SetFillStyle(0);
30
31     // --- Create histograms ---
32     for (Int_t i = 0; i < nHists; i++) {
33         TString name = Form("hist%d", i);
34         hist[i] = new TH1F(name, "", 100, -5, 5);
35         hist[i]->SetMarkerStyle(20 + i);
36         hist[i]->SetMarkerSize(0.8);
37         hist[i]->SetMarkerColor(i + 1);
38         hist[i]->SetLineColor(i + 1);
39         hstack->Add(hist[i], "E1");
40         leg->AddEntry(hist[i], Form("Dataset %d", i + 1), "lep");
41     }
42
43     // --- Generate + Read data ---
44     for (Int_t iFile = 0; iFile < nFiles; iFile++) {
45         TString filename = Form("input%d", iFile);
46
47         // Generate synthetic data with different means/sigmas per
48         // histogram
49         {
50             std::ofstream outfile(filename.Data());
51             if (!outfile.is_open()) {
52                 std::cerr << "Error: cannot write" << filename <<
53                 std::endl;
54                 continue;
55             }

```

```

54
55     for (int ev = 0; ev < nEventsPerFile; ev++) {
56         for (int j = 0; j < nHists; j++) {
57             double val = randGen.Gaus(means[j], sigmas[j]);
58             outfile << val << "\u";
59         }
60         outfile << "\n";
61     }
62     outfile.close();
63 }

64
65 // Read data back into histograms
66 std::ifstream infile(filename.Data());
67 if (!infile.is_open()) {
68     std::cerr << "Warning: could not open " << filename << "\u
69         for reading." << std::endl;
70     continue;
71 }

72 while (true) {
73     for (Int_t j = 0; j < nHists; j++) {
74         Double_t val;
75         infile >> val;
76         if (!infile) break;
77         hist[j]->Fill(val);
78     }
79     if (infile.eof()) break;
80 }
81 infile.close();
82 }

83
84 // --- Fit each histogram & print parameters ---
85 std::cout << "\n===== Gaussian Fit Results =====" << std::endl;
86 TF1 *fitFunc[nHists];
87 for (Int_t i = 0; i < nHists; i++) {
88     fitFunc[i] = new TF1(Form("fit%d", i), "gaus", -5, 5);
89     fitFunc[i]->SetLineColor(hist[i]->GetLineColor());
90     hist[i]->Fit(fitFunc[i], "RQ"); // RQ = quiet fit, but
91         returns result
92
93     double mean = fitFunc[i]->GetParameter(1);
94     double sigma = fitFunc[i]->GetParameter(2);
95     double meanErr = fitFunc[i]->GetParError(1);
96     double sigmaErr = fitFunc[i]->GetParError(2);
97     double chi2ndf = fitFunc[i]->GetChisquare() / fitFunc[i]->
98         GetNDF();
99
100    std::cout << Form("Dataset %d:", i + 1)
101        << "\uMean=\u" << mean << "\u \u" << meanErr
102        << "\uGeV/c,\uSigma=\u" << sigma << "\u \u" <<
103            sigmaErr
104        << "\uGeV/c,\uChi /NDF=\u" << chi2ndf
105        << "\uEntries=\u" << hist[i]->GetEntries()
106        << std::endl;
107
108
109 // --- Draw ---
110 TCanvas *c1 = new TCanvas("c1", "Momentum Distributions", 1000,
111     800);
112 c1->SetGrid();
113 c1->SetTickx();

```

```
110     c1->SetTicky();
111     hstack->Draw("nostack_E1");
112     for (int i = 0; i < nHists; i++) {
113         fitFunc[i]->Draw("same");
114     }
115     hstack->GetXaxis()->SetTitleSize(0.045);
116     hstack->GetYaxis()->SetTitleSize(0.045);
117     hstack->GetXaxis()->SetLabelSize(0.04);
118     hstack->GetYaxis()->SetLabelSize(0.04);
119     leg->Draw();
120
121 // --- Save ---
122 c1->SaveAs("momentum_distributions_with_fits.pdf");
123 c1->SaveAs("momentum_distributions_with_fits.png");
124
125 TFile outFile("momentum_distributions_with_fits.root", "RECREATE");
126     for (Int_t i = 0; i < nHists; i++) {
127         hist[i]->Write();
128         fitFunc[i]->Write();
129     }
130     outFile.Close();
131
132     std::cout << "\nAnalysis complete. Fits overlaid, parameters
133     printed above, ROOT & plots saved." << std::endl;
}
```

4. Analyzing the ρ meson resonance

Listing 4. macro3.c

```

1 void macro3() {
2     gStyle->SetOptStat(0);
3
4     // --- Step 1: Generate simulated signal & background separately
5     // ---
6     int nSignal = 10000;    // 10% of events
7     int nBackground = 90000; // 90% of events
8
9     TRandom3 rand(0);
10
11    // Histograms
12    TH1F* histSignal = new TH1F("histSignal", "Signal\u2225(Breit-Wigner);"
13        Mass\u2225[GeV/c^2];Entries", 100, 0, 2);
14    TH1F* histBackground = new TH1F("histBackground", "Background\u2225(Gaussian);Mass\u2225[GeV/c^2];Entries", 100, 0, 2);
15
16    // Fill signal dataset (Breit-Wigner)
17    for (int i = 0; i < nSignal; ++i) {
18        double mass = rand.BreitWigner(0.77, 0.15);
19        histSignal->Fill(mass);
20    }
21
22    // Fill background dataset (Gaussian)
23    for (int i = 0; i < nBackground; ++i) {
24        double mass = rand.Gaus(0.5, 0.2);
25        histBackground->Fill(mass);
26    }
27
28    // --- Step 2: Apply realistic error bars (stat + syst) ---
29    for (int i = 1; i <= histSignal->GetNbinsX(); ++i) {
30        double N = histSignal->GetBinContent(i);
31        double statErr = sqrt(N);
32        double systErr = 0.05 * N; // 5% systematic
33        histSignal->SetBinError(i, sqrt(statErr*statErr + systErr*systErr));
34    }
35
36    for (int i = 1; i <= histBackground->GetNbinsX(); ++i) {
37        double N = histBackground->GetBinContent(i);
38        double statErr = sqrt(N);
39        double systErr = 0.05 * N; // 5% systematic
40        histBackground->SetBinError(i, sqrt(statErr*statErr + systErr*systErr));
41    }
42
43    // --- Step 3: Fit signal with Breit-Wigner ---
44    TF1* fitBW = new TF1("fitBW", "breitwigner", 0, 2);
45    fitBW->SetParameters(2000, 0.77, 0.15);
46    histSignal->Fit(fitBW, "RQ");
47
48    // --- Step 4: Fit background with Gaussian ---
49    TF1* fitGaus = new TF1("fitGaus", "gaus", 0, 2);
50    fitGaus->SetParameters(8000, 0.5, 0.2);
51    histBackground->Fit(fitGaus, "RQ");
52
53    cout << "Gaussian\u2225Chi2/NDF\u2225(background):" << fitGaus->
54        GetChisquare()/fitGaus->GetNDF() << endl;

```

```

51     cout << "Breit-Wigner\u2022Chi2/NDF\u2022(signal):" << fitBW->GetChisquare
52     ()/fitBW->GetNDF() << endl;
53
54 // --- Step 5: Plot results ---
55 TCanvas* c = new TCanvas("c", "Invariant\u2022Mass\u2022Analysis", 900,
56                           700);
57
58 histBackground->SetMarkerStyle(20);
59 histBackground->SetMarkerColor(kBlue);
60 histBackground->SetLineColor(kBlue);
61
62 histSignal->SetMarkerStyle(21);
63 histSignal->SetMarkerColor(kRed);
64 histSignal->SetLineColor(kRed);
65
66 histBackground->Draw("E1");
67 histSignal->Draw("E1\u2022SAME");
68
69 fitGaus->SetLineColor(kBlue+2);
70 fitGaus->SetLineWidth(2);
71 fitBW->SetLineColor(kRed+2);
72 fitBW->SetLineWidth(2);
73
74 TLegend* leg = new TLegend(0.6, 0.7, 0.88, 0.88);
75 leg->SetBorderSize(0);
76 leg->AddEntry(histBackground, "Background\u2022(Gaussian)", "lep");
77 leg->AddEntry(histSignal, "Signal\u2022(Breit-Wigner)", "lep");
78 leg->AddEntry(fitGaus, "Gaussian\u2022Fit", "l");
79 leg->AddEntry(fitBW, "Breit-Wigner\u2022Fit", "l");
80 leg->Draw();
81
82 // --- Step 6: Save outputs ---
83 c->SaveAs("signal_background.pdf");
84 c->SaveAs("signal_background.png");
85
86 TFile outFile("analysis_output.root", "RECREATE");
87 histSignal->Write();
88 histBackground->Write();
89 outFile.Close();
90
91 cout << "Analysis\u2022complete.\u2022Separate\u2022signal\u2022and\u2022background\u2022stored
92 ."
93     << endl;
94 }
```

5. Creating a 2D histogram

Listing 5. *two_dhistogram.c*

```

1 void two_d_histogram() {
2     // --- Settings ---
3     const int nEvents = 1e6;
4     const double meanVal = 0.0;
5     const double sigmaVal = 1.0;
6
7     // --- Create histograms ---
8     TH2F *h2 = new TH2F("h2", ";x\u209c[cm];y\u209c[cm]", 200, -3, 3, 200, -3,
9                         3);
10
11    // Random generator
12    TRandom3 rand(10);
13
14    // Fill 2D Gaussian
15    for (int i = 0; i < nEvents; i++) {
16        double x = rand.Gaus(meanVal, sigmaVal);
17        double y = rand.Gaus(meanVal, sigmaVal);
18        h2->Fill(x, y);
19    }
20
21    // --- Create canvas with pads for projections ---
22    TCanvas *c = new TCanvas("c", "STAR-style\u209c2D\u209cGaussian\u209cwith\u209c
23        Projections", 1000, 1000);
24    gStyle->SetOptStat(0);
25    gStyle->SetPalette(kBird);
26    gStyle->SetNumberContours(100);
27
28    // Pads
29    TPad *padMain = new TPad("padMain", "Main\u209c2D", 0.0, 0.0, 0.8,
30                             0.8);
31    TPad *padX      = new TPad("padX", "X\u209cprojection", 0.0, 0.8, 0.8,
32                             1.0);
33    TPad *padY      = new TPad("padY", "Y\u209cprojection", 0.8, 0.0, 1.0,
34                             0.8);
35
36    padMain->SetRightMargin(0.15);
37    padMain->SetLeftMargin(0.15);
38    padMain->SetBottomMargin(0.15);
39    padX->SetBottomMargin(0.0);
40    padX->SetLeftMargin(0.15);
41    padX->SetRightMargin(0.15);
42    padY->SetTopMargin(0.15);
43    padY->SetLeftMargin(0.0);
44
45    padMain->Draw();
46    padX->Draw();
47    padY->Draw();
48
49    // --- Draw main 2D plot ---
50    padMain->cd();
51    h2->Draw("COLZ");
52    h2->Draw("CONT3\u209cSAME"); // add contour lines on top
53
54    // --- X projection ---
55    padX->cd();
56    TH1D *hX = h2->ProjectionX();

```

```

52     hX->SetLineColor(kBlue+1);
53     hX->SetFillColorAlpha(kBlue, 0.3);
54     hX->GetXaxis()->SetLabelSize(0);
55     hX->GetYaxis()->SetTitle("Entries");
56     hX->Draw();
57
58 // --- Y projection ---
59 padY->cd();
60 TH1D *hY = h2->ProjectionY();
61 hY->SetLineColor(kRed+1);
62 hY->SetFillColorAlpha(kRed, 0.3);
63 hY->GetYaxis()->SetLabelSize(0);
64 hY->Draw("HIST");
65 hY->SetMaximum(hY->GetMaximum() * 1.1);
66
67 // --- Output statistics to terminal ---
68 double meanX = h2->GetMean(1);
69 double meanY = h2->GetMean(2);
70 double sigmaX = h2->GetRMS(1);
71 double sigmaY = h2->GetRMS(2);
72
73 std::cout << "=====STAR-style 2D Gaussian Analysis=====" << std::endl;
74 std::cout << "Number of entries:" << h2->GetEntries() << std::endl;
75 std::cout << "Mean X = " << meanX << " cm" << std::endl;
76 std::cout << "Sigma X = " << sigmaX << " cm" << std::endl;
77 std::cout << "Mean Y = " << meanY << " cm" << std::endl;
78 std::cout << "Sigma Y = " << sigmaY << " cm" << std::endl;
79
80 // --- Save outputs ---
81 c->SaveAs("beamspot_2D_with_projections.pdf");
82 c->SaveAs("beamspot_2D_with_projections.png");
83
84 TFile outFile("beamspot_2D_with_projections.root", "RECREATE");
85 h2->Write();
86 hX->Write("projX");
87 hY->Write("projY");
88 outFile.Close();
89 }
```

6. Creating a 3D visual

Listing 6. view23d.c

```

1 void view2_3d() {
2     gStyle->SetOptStat(0);
3
4     // --- Canvas ---
5     TCanvas *c1 = new TCanvas("c1", "STAR-style 3D Geometry", 1200,
6                               900);
7
8     // --- Geometry Manager ---
9     TGeoManager *man = new TGeoManager("STARgeom", "STAR Detector"
10                                         "Geometry");
11
12     // --- Materials and Mediums ---
13     TGeoMaterial *matVacuum = new TGeoMaterial("Vacuum", 0, 0, 0);
14     TGeoMedium   *vacuum    = new TGeoMedium("Vacuum", 1, matVacuum);
15
16     TGeoMaterial *matAl = new TGeoMaterial("Aluminum", 26.98, 13,
17                                           2.7);
18     TGeoMedium   *alum   = new TGeoMedium("Aluminum", 2, matAl);
19
20     TGeoMaterial *matSi = new TGeoMaterial("Silicon", 28.085, 14,
21                                           2.33);
22     TGeoMedium   *silicon = new TGeoMedium("Silicon", 3, matSi);
23
24     // --- World volume ---
25     TGeoVolume *top = man->MakeBox("TOP", vacuum, 50, 50, 50);
26     man->SetTopVolume(top);
27
28     // --- Barrel detector (tube) ---
29     TGeoVolume *barrel = man->MakeTube("TPC", silicon, 20.0, 22.0,
30                                         40.0);
31     barrel->SetLineColor(kBlue - 7);
32     barrel->SetTransparency(60);
33     top->AddNode(barrel, 0);
34
35     // --- Endcap disk ---
36     TGeoVolume *endcap = man->MakeTube("Endcap", alum, 0.0, 22.0,
37                                         1.0);
38     endcap->SetLineColor(kGreen + 2);
39     endcap->SetTransparency(40);
40     top->AddNode(endcap, 0, new TGeoTranslation(0, 0, 40.0));
41     top->AddNode(endcap, 1, new TGeoTranslation(0, 0, -40.0));
42
43     // --- Additional inner tracker layer ---
44     TGeoVolume *innerLayer = man->MakeTube("InnerTracker", silicon,
45                                             5.0, 5.5, 40.0);
46     innerLayer->SetLineColor(kOrange + 7);
47     innerLayer->SetTransparency(40);
48     top->AddNode(innerLayer, 0);
49
50     // --- Close geometry ---
51     man->CloseGeometry();
52
53     // --- Draw geometry ---
54     top->Draw("ogl");
55
56     // --- Access 3D viewer ---
57
58 }
```

```

50     TGLViewer *view = (TGLViewer*)gPad->GetViewer3D();
51     if (view) {
52         view->SetStyle(TGLRnrCtx::kOutline); // Outline mode for
53         clarity
54         view->CurrentCamera().RotateRad(-0.5, 0.5); // nice angle
55         view->RequestDraw();
56     }
57
58     // --- Add multiple tracks ---
59     TRandom3 r(0);
60     for (int t = 0; t < 5; t++) {
61         TPolyLine3D *track = new TPolyLine3D();
62         track->SetLineColor(kRed + (t % 3));
63         track->SetLineWidth(2);
64         double phi0 = r.Uniform(0, 2 * TMath::Pi());
65         double radius = r.Uniform(5.0, 20.0);
66         double zpos = -40.0;
67         for (int p = 0; p < 100; p++) {
68             double angle = phi0 + p * 0.02;
69             double x = radius * cos(angle);
70             double y = radius * sin(angle);
71             double z = zpos + p * (80.0 / 100);
72             track->SetPoint(p, x, y, z);
73         }
74         track->Draw("same");
75     }
76
77     // --- Annotation ---
78     TLatex label;
79     label.SetNDC();
80     label.SetTextSize(0.035);
81     label.DrawLatex(0.02, 0.96, "#bf{STAR\u201dDetector\u201dGeometry\u201d\u201d"
82     "Simulation}");  

83     label.DrawLatex(0.02, 0.92, "TPC\u201d+Endcaps\u201d+Inner\u201dTracker");
84
85     // --- Legend for materials ---
86     TLegend *leg = new TLegend(0.7, 0.7, 0.9, 0.88);
87     leg->AddEntry(barrel, "TPC\u201dBarrel\u201d(Si)", "f");
88     leg->AddEntry(endcap, "Endcaps\u201d(A1)", "f");
89     leg->AddEntry(innerLayer, "Inner\u201dTracker\u201d(Si)", "f");
90     leg->Draw();
91
92     // --- Save outputs ---
93     c1->SaveAs("star_geometry_full.png");
94     c1->SaveAs("star_geometry_full.pdf");
95
96     std::cout << "3D\u201dSTAR-style\u201dgeometry\u201dsaved\u201das\u201dPNG\u201dand\u201dPDF.\n";
}

```

7. Proton-Proton collisions

Listing 7. pccollision.cc

```

1 #include "Pythia8/Pythia.h"
2 #include "TFile.h"
3 #include "TH1F.h"
4 #include "TH2F.h"
5 #include "TCanvas.h"
6 #include "TLegend.h"
7 #include "TStyle.h"
8 #include <cmath>
9
10 int main() {
11     // --- Settings ---
12     int nevents = 5000;      // Number of events
13     double eCM = 200.0;       // RHIC energy (GeV) for STAR pp
14     collisions
15
16     // --- Initialize Pythia ---
17     Pythia8::Pythia pythia;
18     pythia.readString("Beams:idA=2212");
19     pythia.readString("Beams:idB=2212");
20     pythia.readString("Beams:eCM=200."); // STAR pp energy
21     //pythia.readString("SoftQCD:inelastic = on");
22     pythia.readString("HardQCD:all=on");
23     pythia.init();
24
25     // --- ROOT histograms ---
26     TH1F *h_pT = new TH1F("h_pT", ";p_{T}[GeV/c];Entries", 100, 0,
27                           5);
28     TH1F *h_eta = new TH1F("h_eta", "#eta;Entries", 100, -5, 5);
29     TH1F *h_phi = new TH1F("h_phi", "#phi[rad];Entries", 64, -M_PI,
30                           M_PI);
31     TH2F *h_pT_eta = new TH2F("h_pT_eta", "#eta;p_{T}[GeV/c]", 50,
32                               -2.5, 2.5, 50, 0, 5);
33
34     h_pT->Sumw2();
35     h_eta->Sumw2();
36     h_phi->Sumw2();
37     h_pT_eta->Sumw2();
38
39     // --- Event loop ---
40     for (int i = 0; i < nevents; i++) {
41         if (!pythia.next()) continue;
42
43         for (int j = 0; j < pythia.event.size(); j++) {
44             if (!pythia.event[j].isFinal()) continue; // Final state
45             only
46
47             double pT = pythia.event[j].pT();
48             double eta = pythia.event[j].eta();
49             double phi = pythia.event[j].phi();
50
51             // STAR acceptance cut
52             if (fabs(eta) < 1.0 && pT > 0.2) {
53                 h_pT->Fill(pT);
54                 h_eta->Fill(eta);
55                 h_phi->Fill(phi);
56             }
57         }
58     }
59 }
```

```

52
53         // Fill 2D histogram without acceptance cut to show full
54         // coverage
55         h_pT_eta->Fill(eta, pT);
56     }
57
58     // --- Save histograms to ROOT file ---
59     TFile outFile("pythia_histograms.root", "RECREATE");
60     h_pT->Write();
61     h_eta->Write();
62     h_phi->Write();
63     h_pT_eta->Write();
64     outFile.Close();
65
66     // --- STAR-style plotting ---
67     gStyle->SetOptStat(0);
68     gStyle->SetTitleFontSize(0.05);
69
70     // pT plot
71     TCanvas *c1 = new TCanvas("c1", "pT_Distribution", 800, 600);
72     c1->SetLogy();
73     h_pT->SetMarkerStyle(20);
74     h_pT->Draw("E1");
75     c1->SaveAs("pT_distribution.pdf");
76     c1->SaveAs("pT_distribution.png");
77
78     // eta plot
79     TCanvas *c2 = new TCanvas("c2", "Eta_Distribution", 800, 600);
80     h_eta->SetMarkerStyle(20);
81     h_eta->Draw("E1");
82     c2->SaveAs("eta_distribution.pdf");
83     c2->SaveAs("eta_distribution.png");
84
85     // phi plot
86     TCanvas *c3 = new TCanvas("c3", "Phi_Distribution", 800, 600);
87     h_phi->SetMarkerStyle(20);
88     h_phi->Draw("E1");
89     c3->SaveAs("phi_distribution.pdf");
90     c3->SaveAs("phi_distribution.png");
91
92     // 2D pT vs eta plot
93     TCanvas *c4 = new TCanvas("c4", "pT_vs_Eta", 900, 700);
94     c4->SetRightMargin(0.15);
95     gStyle->SetPalette(kBird);
96     h_pT_eta->Draw("COLZ");
97     c4->SaveAs("pT_vs_eta.pdf");
98     c4->SaveAs("pT_vs_eta.png");
99
100    // --- Print Pythia statistics ---
101    pythia.stat();
102
103    return 0;
104 }
```