

A Search for Light Axion-Like Particles: Using $Z \rightarrow ee$ to Validate Transformer-Based Identification of Boosted Diphotons

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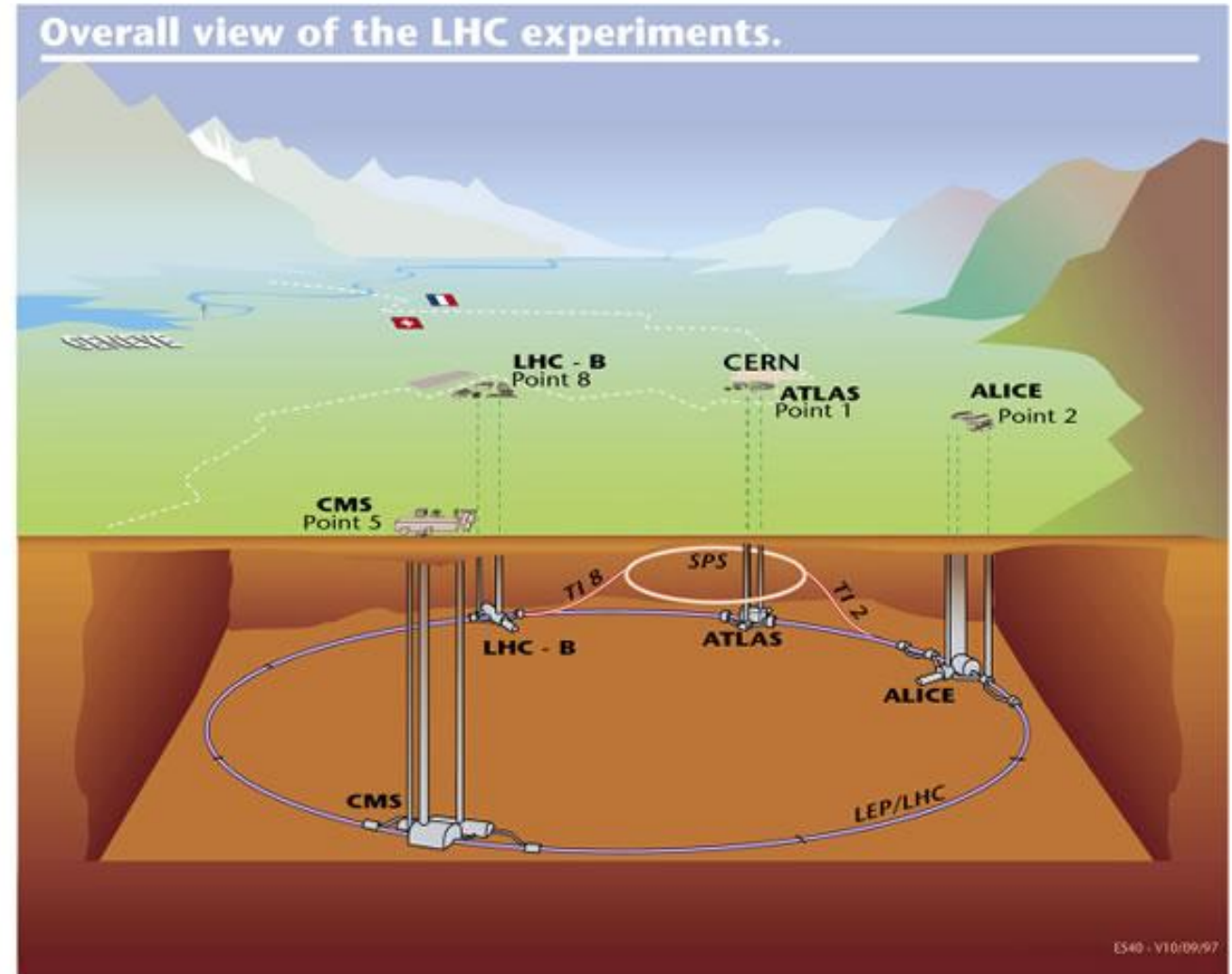


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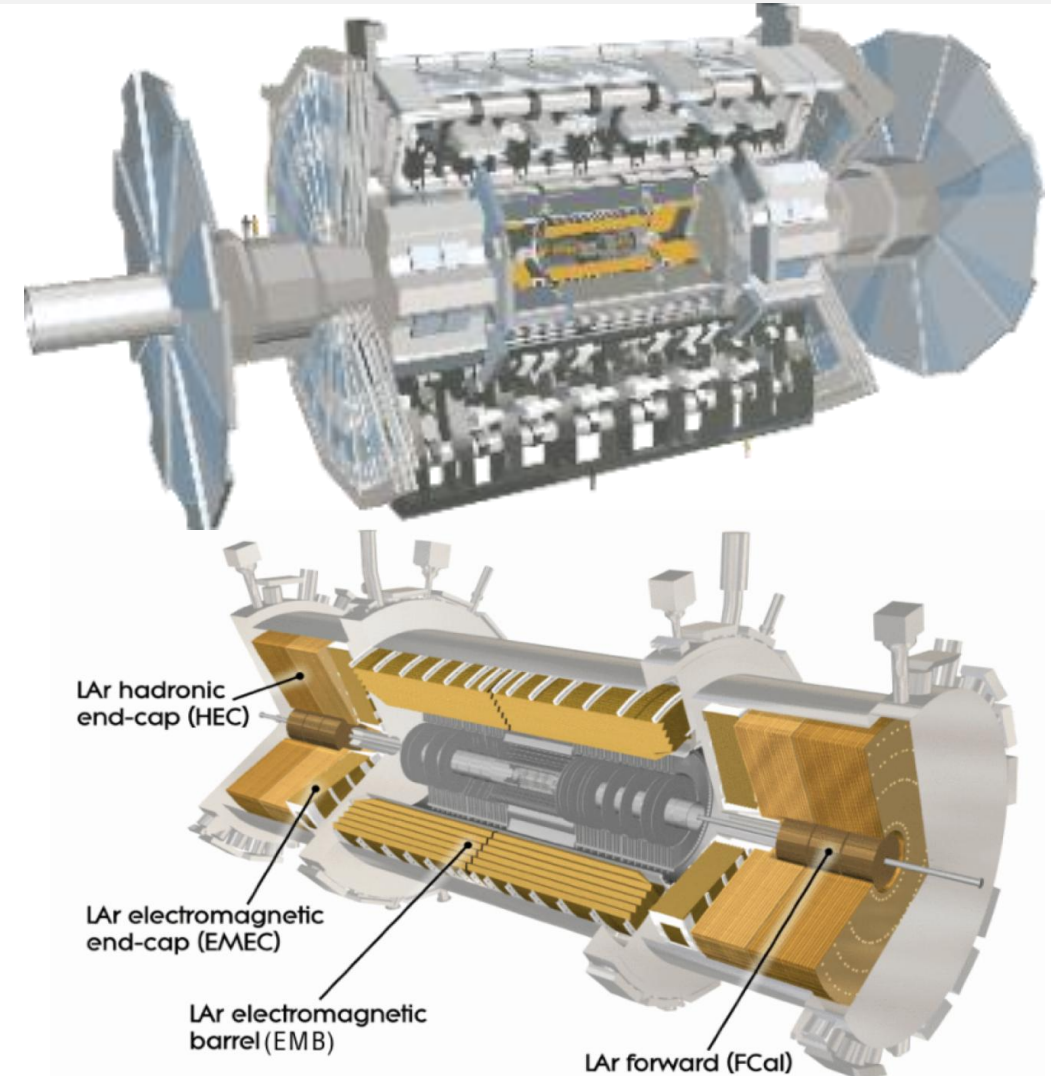
The Large Hadron Collider

- Large Hadron Collider (LHC): The most powerful particle accelerator in the world
- Particle collisions at energies of $\sqrt{s} = 14 \text{ TeV}$
- Four major collision points: ATLAS, CMS, ALICE, LHCb



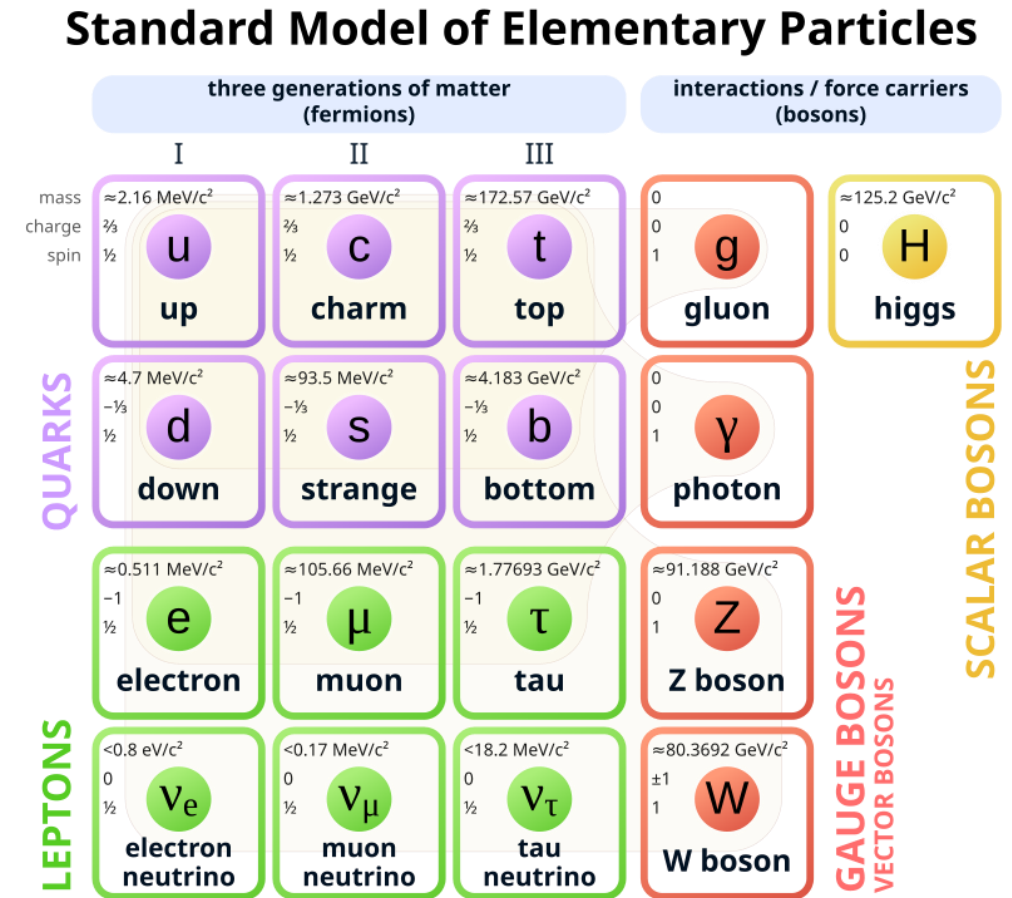
The ATLAS Detector

- A Toroidal LHC ApparatuS (ATLAS): the largest detector at the LHC
- 3 Concentric Cylindrical Layers:
 - Inner Detector
 - Calorimeters (EM and hadronic)
 - Muon spectrometer



Addressing the CP Problem: Axion-Like Particles (ALPs)

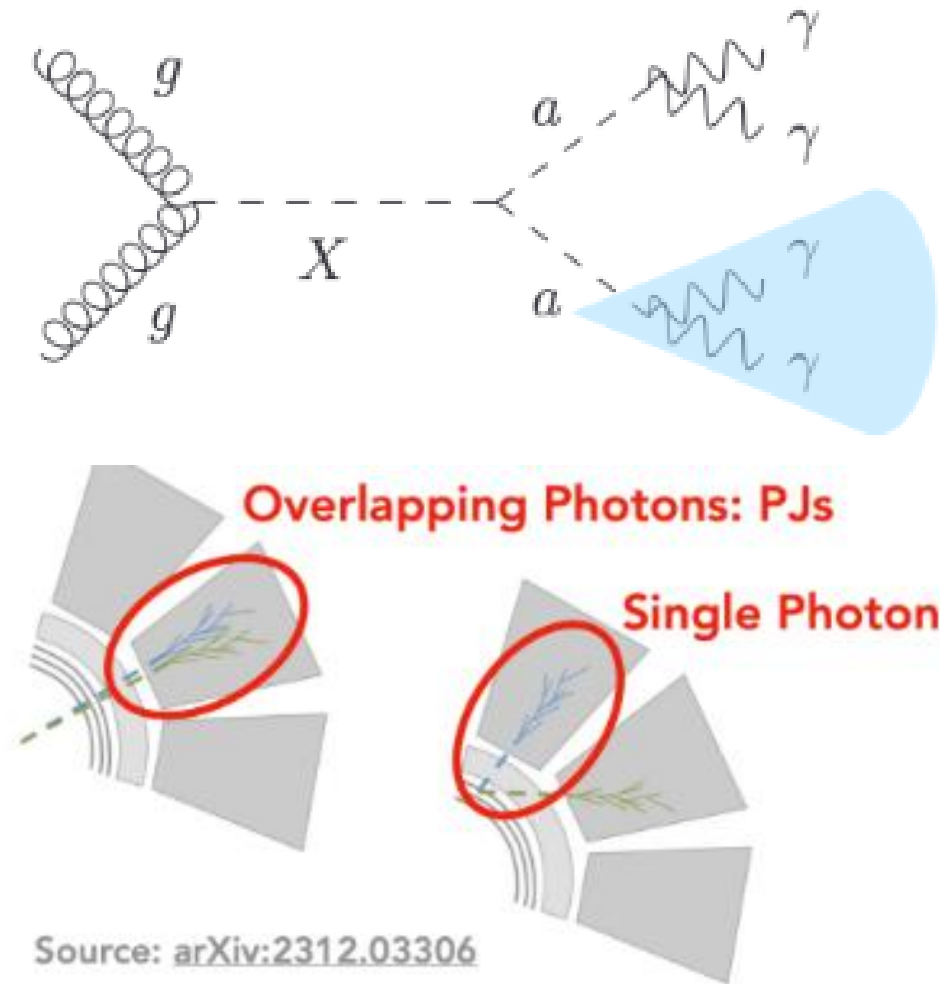
- A conflict in the Standard Model: The strong force does not break CP symmetry
- Theoretical axions offer a solution to the CP problem
- ALPs share properties with axions and are candidates for dark matter
- Columbia's search for ALPs: $X \rightarrow aa \rightarrow 4\gamma$



Machine Learning-Based Photon Classification

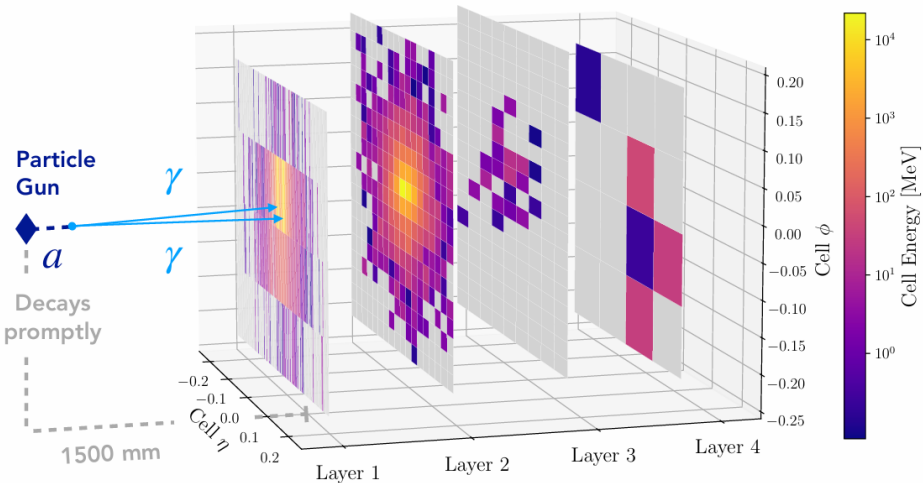
- The signal $X \rightarrow aa \rightarrow 4\gamma$ decay produces boosted photon pairs that hit the detector very close together
- Merged photon pairs (signal) and single photons (background) are nearly identical in the detector
- The Columbia analysis uses a **transformer** trained to distinguish merged photon pairs from single photons

Transformer: [arXiv:2607.08175](https://arxiv.org/abs/2607.08175) [hep-ph]



Cell-Level Information Optimizes Model Performance

- The ATLAS calorimeter is composed of small cells that measure particle energy
- Shower shape variables (SSVs) characterize many-cell regions using a few variables
- Using **full cell-level information** to train the transformer **improves performance** over previous SSV-trained classifiers

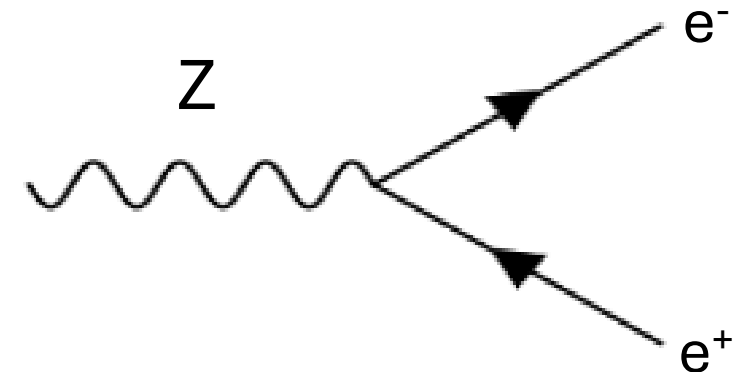
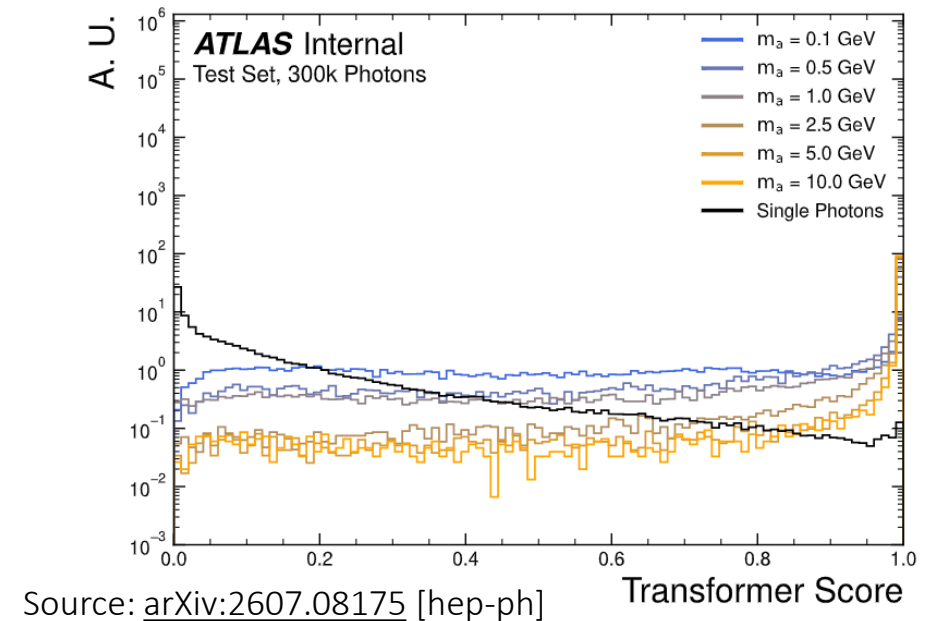


		Background Rejection ($1/\epsilon_{\text{bkg}}$)		
Model	Overall AUC	90% ϵ_{sig}	95% ϵ_{sig}	99% ϵ_{sig}
SSV-Level				
BDT	0.92	3.2	1.8	1.2
DNN	0.92	3.1	1.8	1.2
Cell-Level				
CNN	0.95	6.4	2.6	1.3
PFN	0.97	62 ± 7	5.4 ± 0.2	1.48 ± 0.01
Transformer	0.98	472 ± 60	9.0 ± 0.4	1.64 ± 0.01
MLP Mixer	0.93	4.3	2.1	1.2

Source: [arXiv:2607.08175](https://arxiv.org/abs/2607.08175) [hep-ph]

Validating the Transformer with Data

- Initial training with undiscovered particles can only be evaluated on simulated samples
- To compare the model's performances on simulation and on data, the transformer can be evaluated with **background samples**
- $Z \rightarrow e^+ e^-$ decay produces **electrons**, which look very similar to **single photons** in the calorimeter



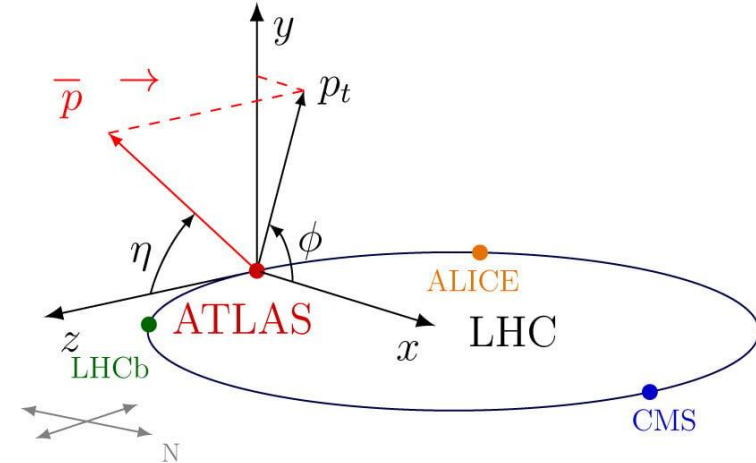
$Z \rightarrow e^+e^-$ Sample Composition

$Z \rightarrow e^+e^-$ Sample from 2022 (Run 3)

Data: 19,478,724 events

Simulation: 37,320,526 events

- The simulation aims to perfectly mimic the data
- Ideally, the transformer should perform equally on simulation and data samples

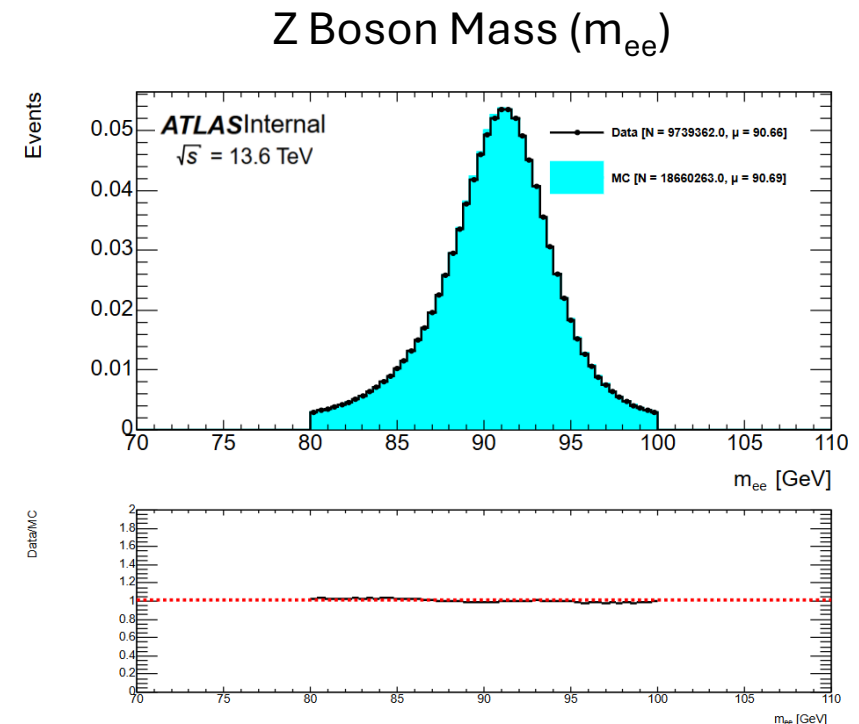
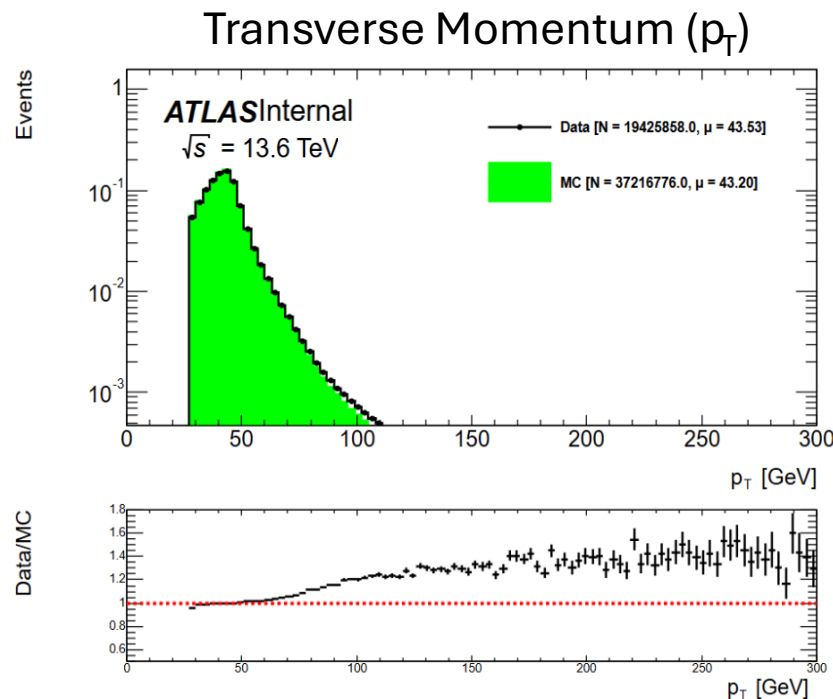


Cuts Applied to $Z \rightarrow e^+e^-$ Background Sample

Value	Cut
Transverse Momentum (p_T)	$p_T > 20 \text{ GeV}$
Z Boson Mass (m_{ee})	$80 \text{ GeV} < m_{ee} < 100 \text{ GeV}$
Pseudorapidity (η)	$ \eta < 1.37, 1.52 < \eta < 2.5$
Net Electron Pair Charge	Net Charge = 0

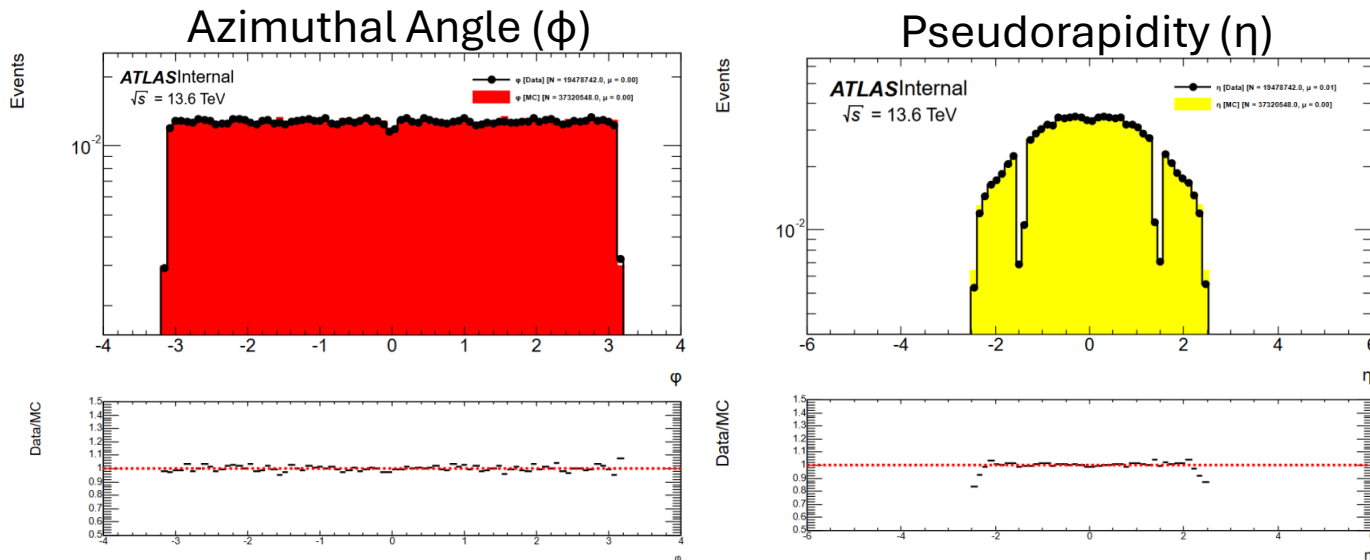
p_T and Z Boson Mass in the Sample

- Good agreement between simulation and data in both p_T and m_{ee}
- Z boson mass averages to ~ 90.6 GeV for data and simulation (compare to 91.2 GeV)

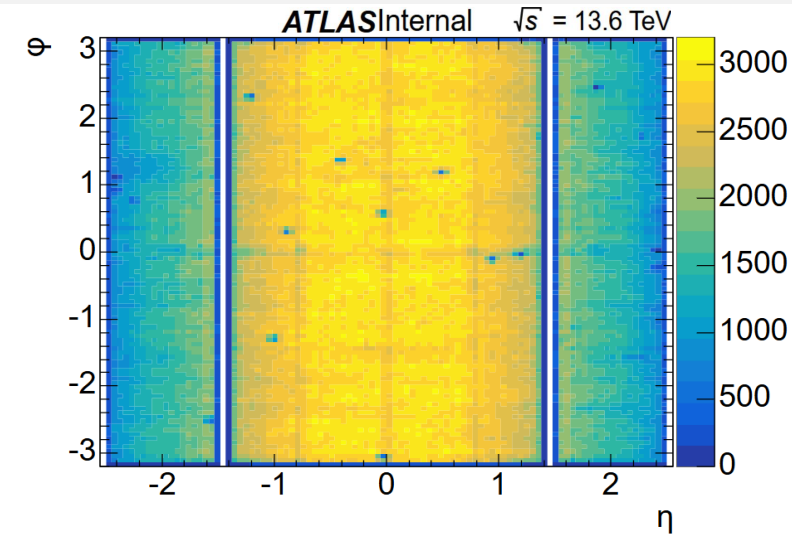


Spatial Distribution of Events in the Detector

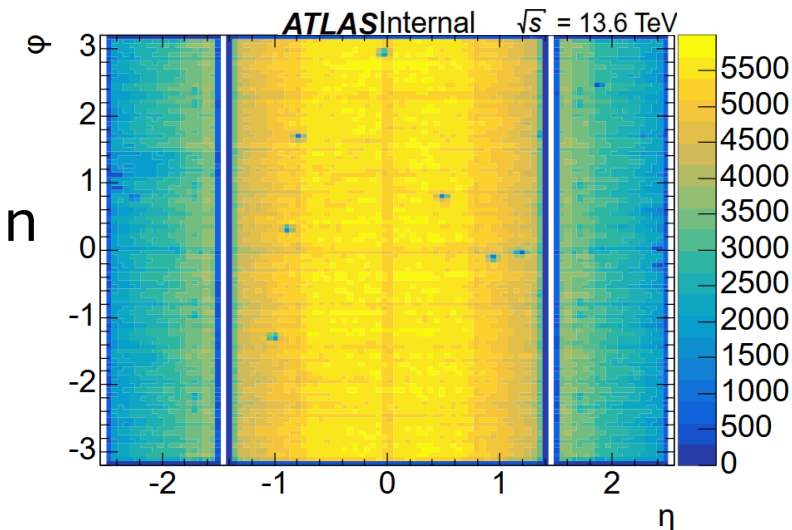
- Good agreement between simulation and data in both η and ϕ
- Unrolling the inner surface of the detector around ϕ creates an η - ϕ plane



Data



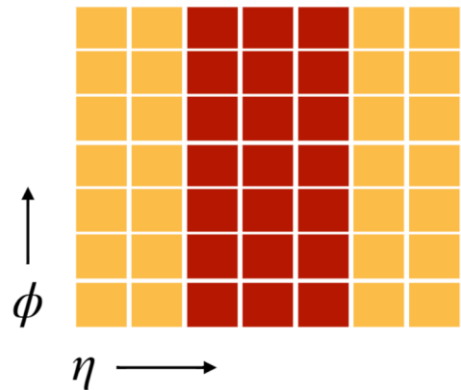
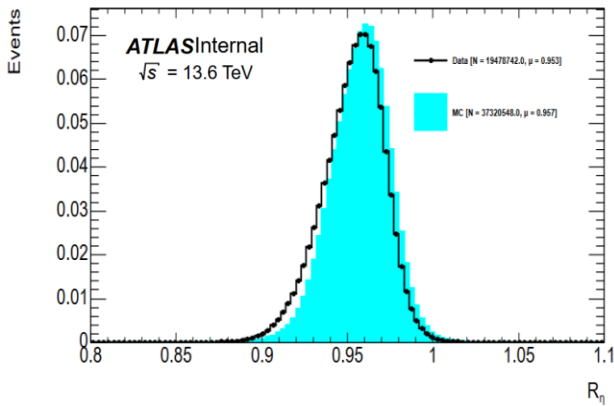
Simulation



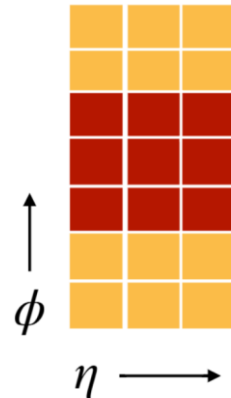
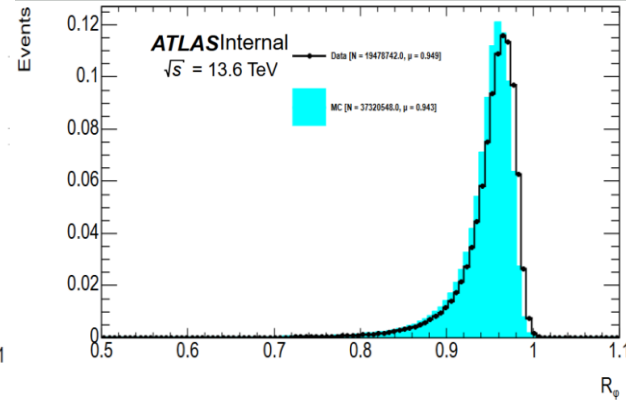
Shower Shape Variable Agreement Between Data and Simulation

Data and simulation show reasonable agreement across shower shape variables

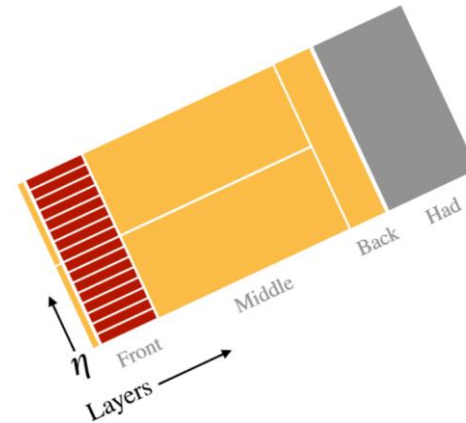
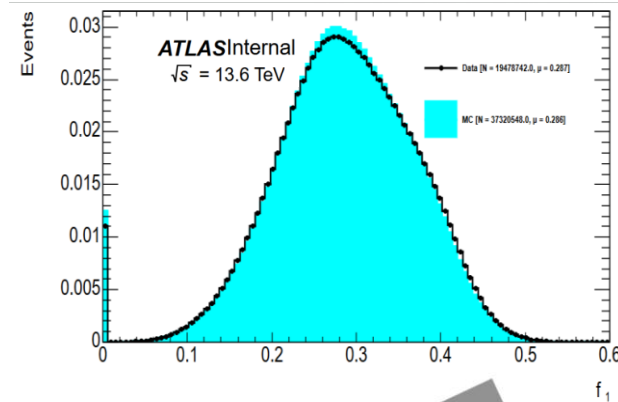
η Energy Ratio (R_η)



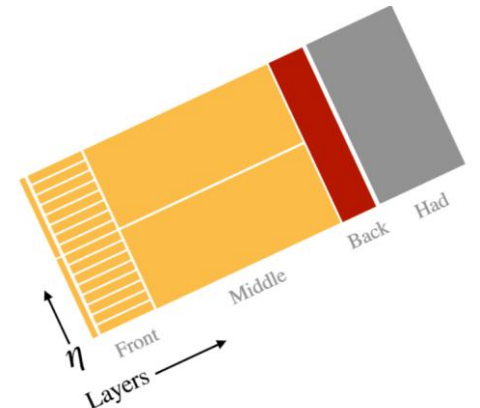
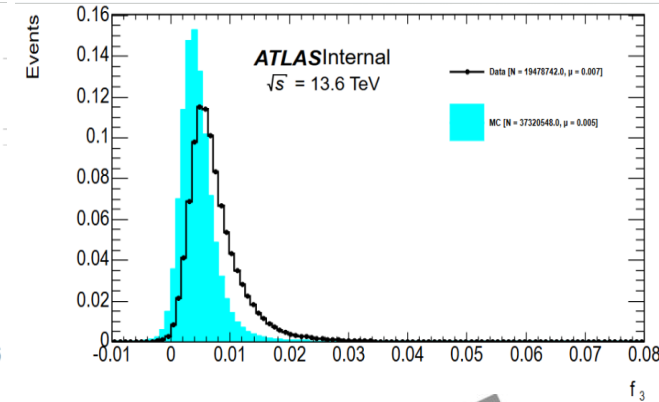
ϕ Energy Ratio (R_ϕ)



Energy Fraction (f_1)



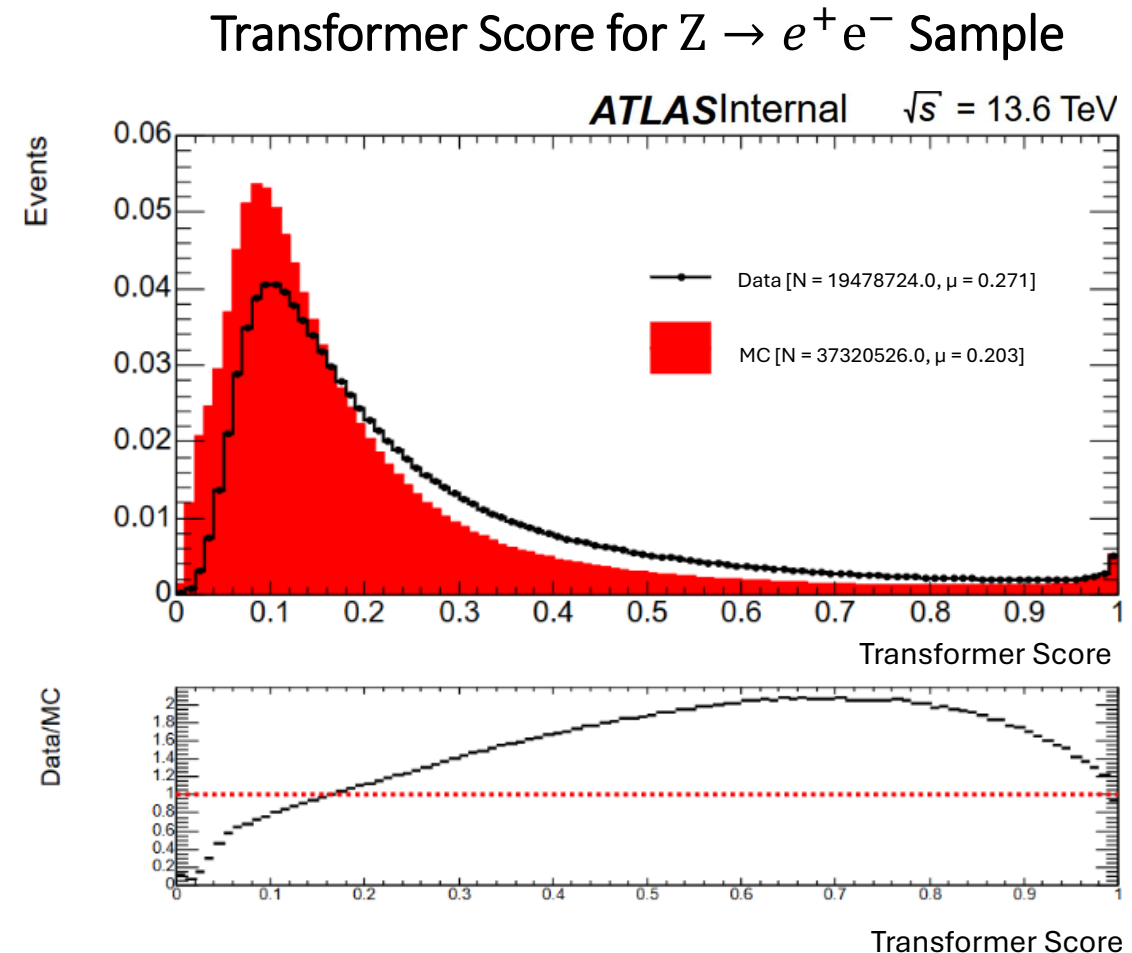
Energy Fraction (f_3)



Data-Simulation Disagreement in the Transformer

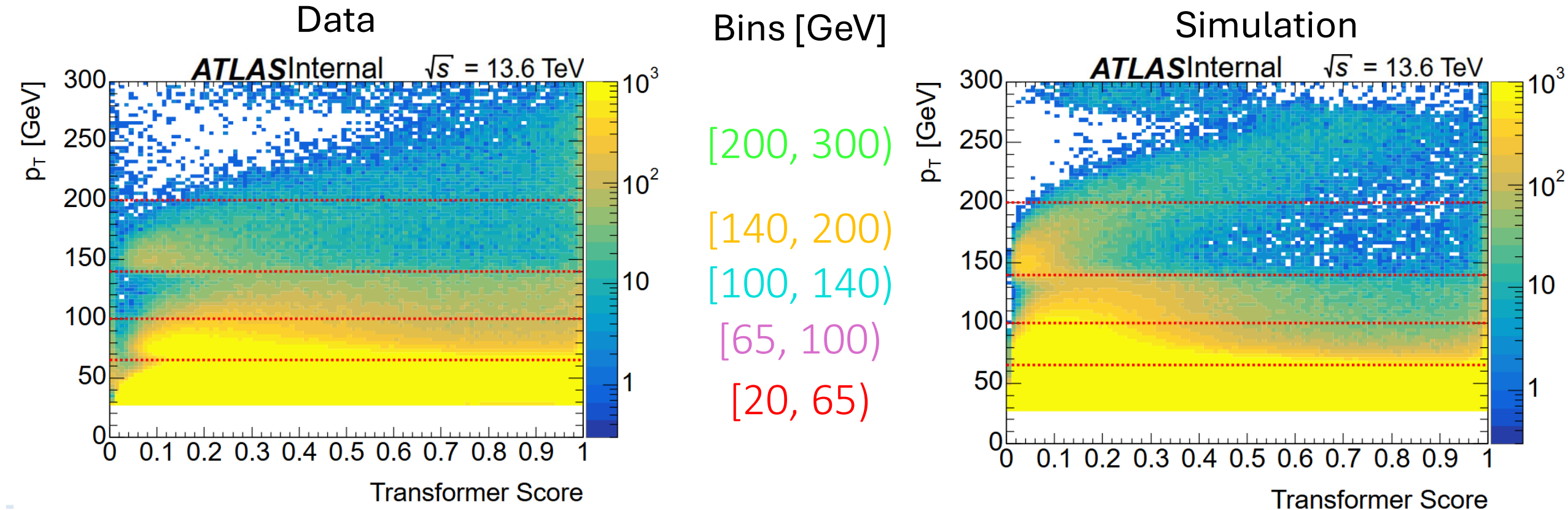
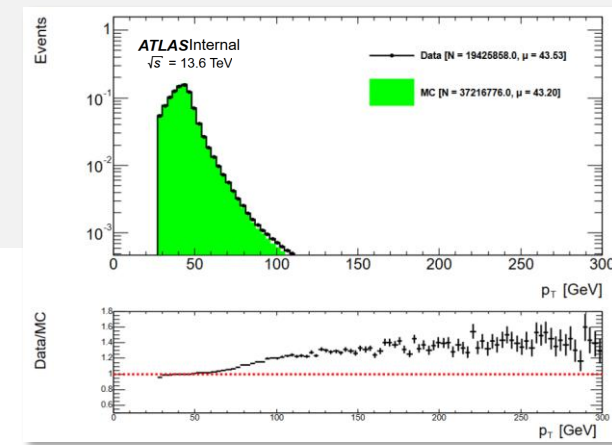
Unexpectedly, the transformer performance on electrons is not the same in data and simulation

- As expected, the transformer categorizes the electrons as background-like
- Can correlations with physical values like kinematics or shower shape variables reveal an underlying pattern behind the transformer's inconsistency?



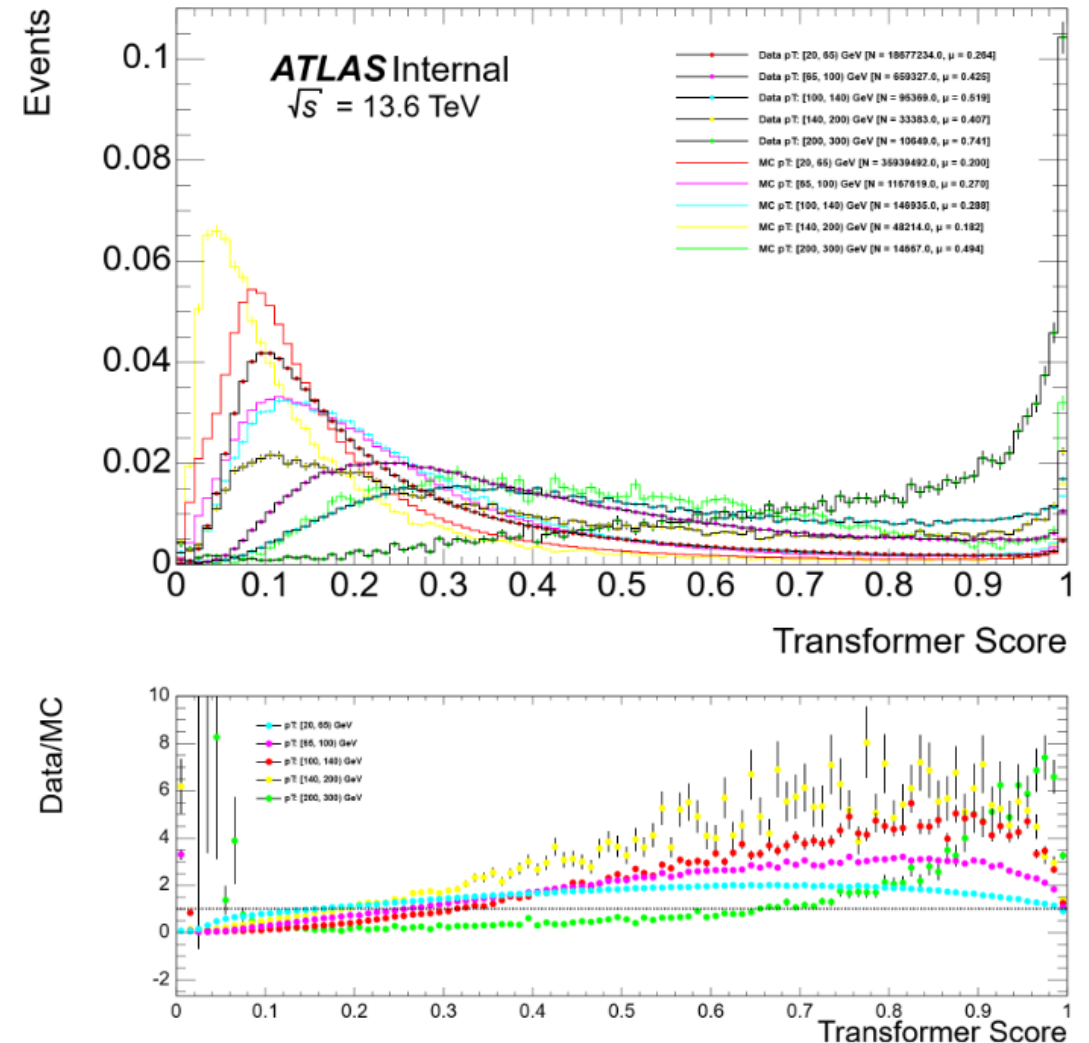
p_T Bins According to Transformer Score Distribution

- There is noticeable variation in transformer score across p_T



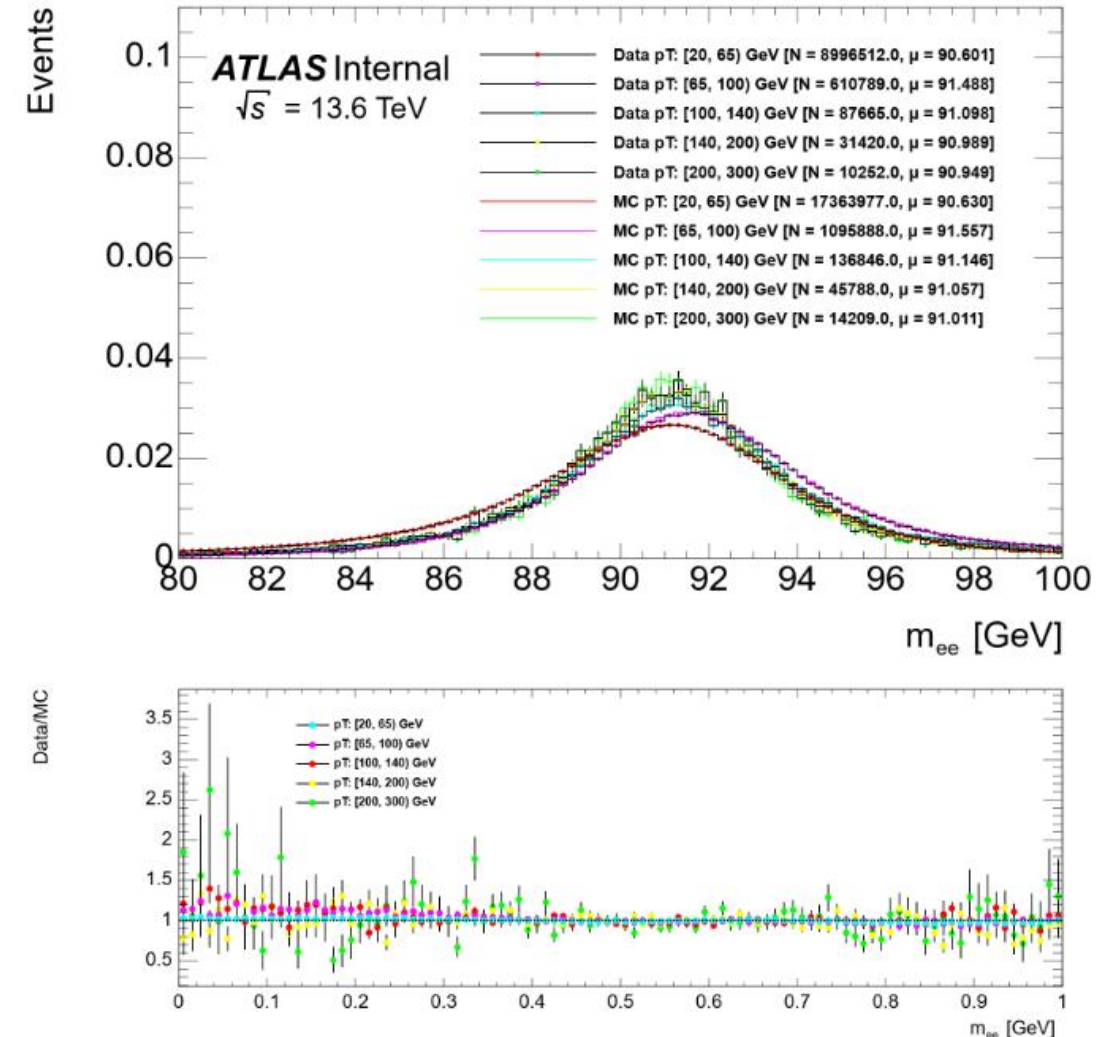
p_T -Binned Transformer Score

- p_T Bins (GeV): [20, 65), [65, 100), [100, 140), [140, 200), [200, 300)
- There is a strong disagreement between data and simulation in the highest p_T bin
- Highest p_T bin $\approx 0.1\%$ of overall sample



Sample Purity Across p_T Bins

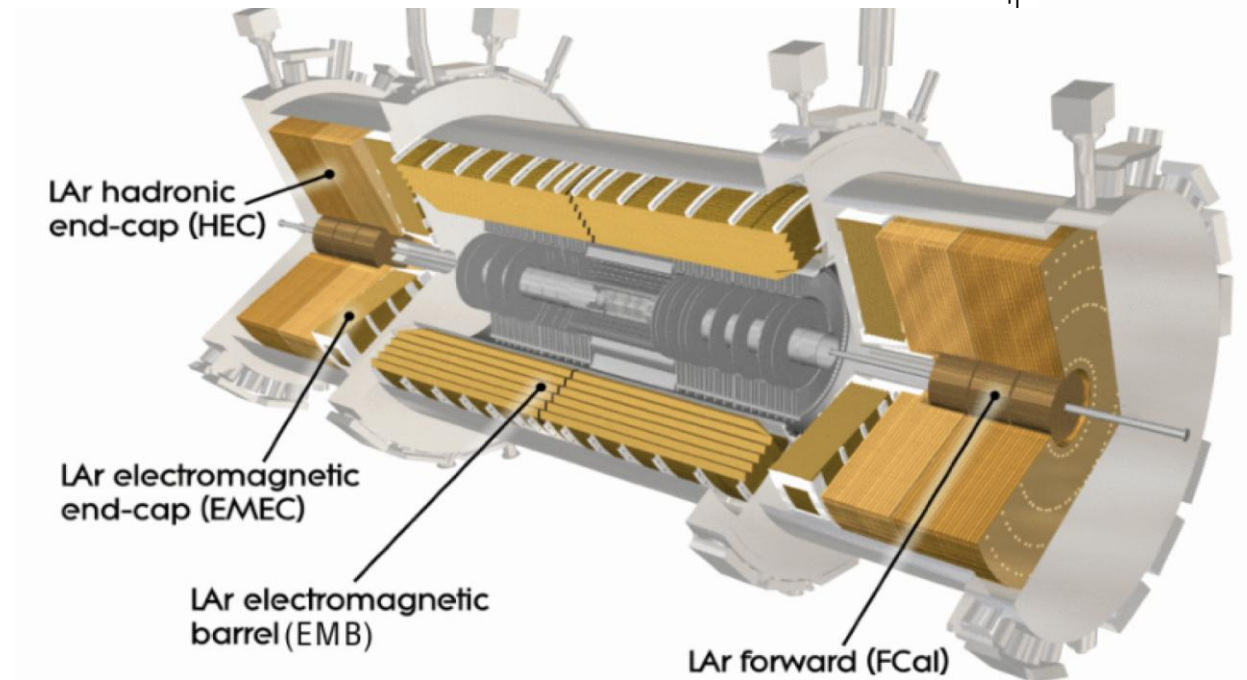
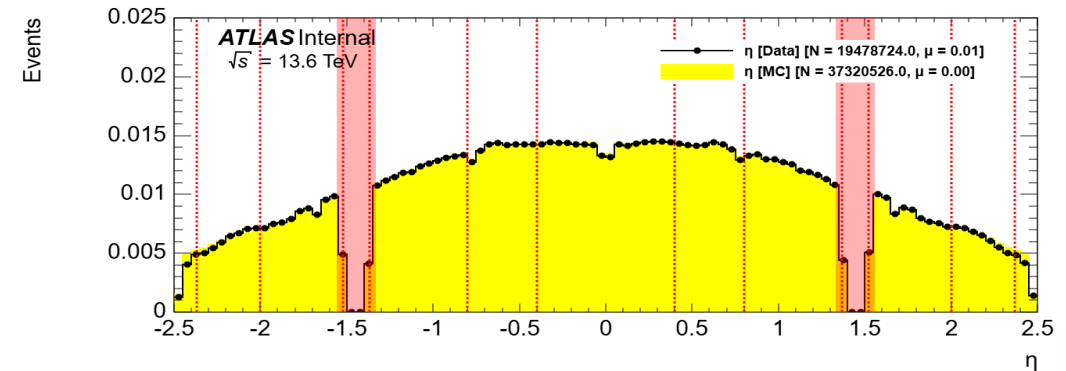
- Electron parent mass shows no significant variation across p_T bins
- All bins average between 90.60 GeV and 91.55 GeV (Z boson: 91.2 GeV)
- The data-simulation disagreement is not due to lack of sample purity at high p_T



η Bins According to Detector Geometry

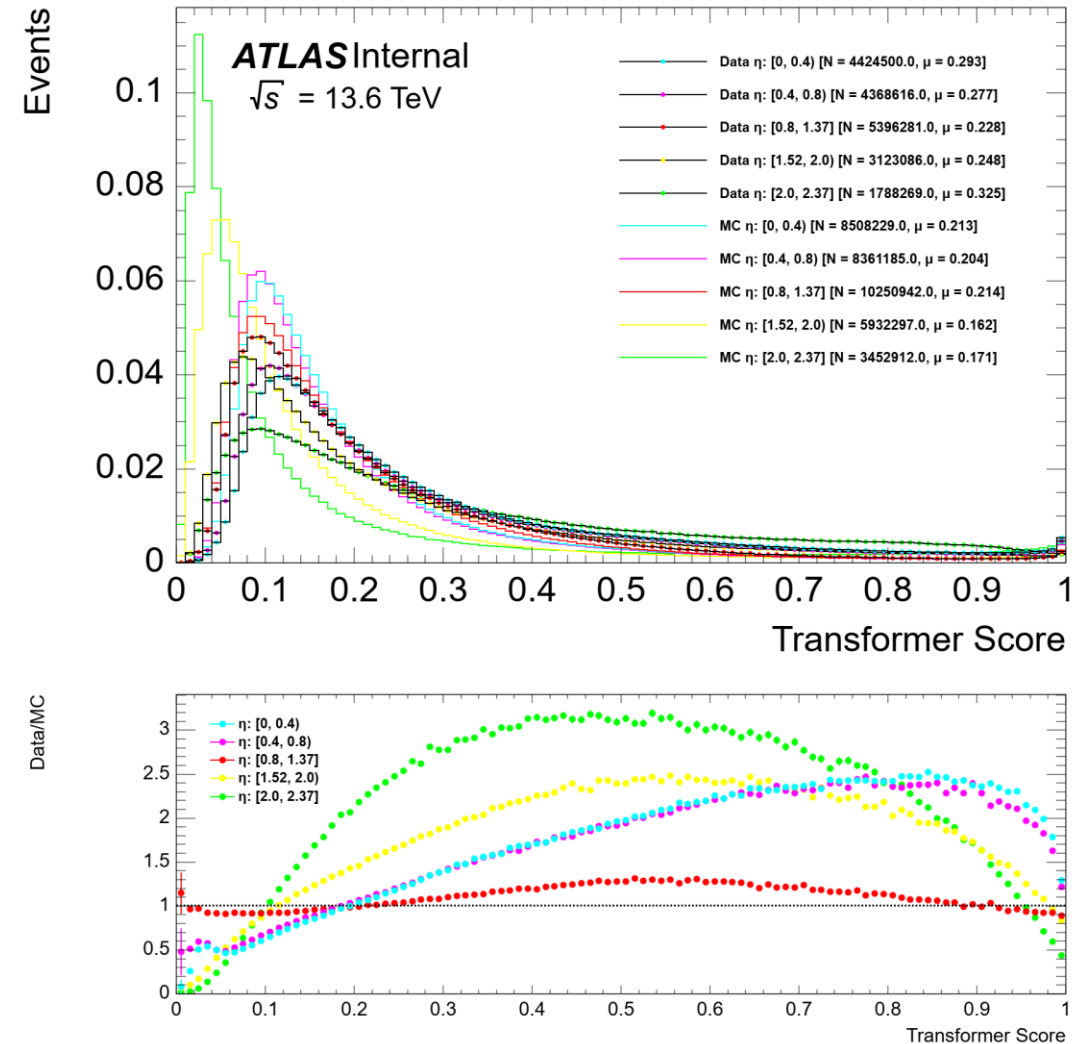
Different regions of the calorimeter correspond to different values of $|\eta|$

- Inner Barrel: $[0, 0.4)$, $[0.4, 0.8)$
- Outer Barrel: $[0.8, 1.37)$
- Endcap: $[1.52, 2.0)$, $[2.0, 2.37)$



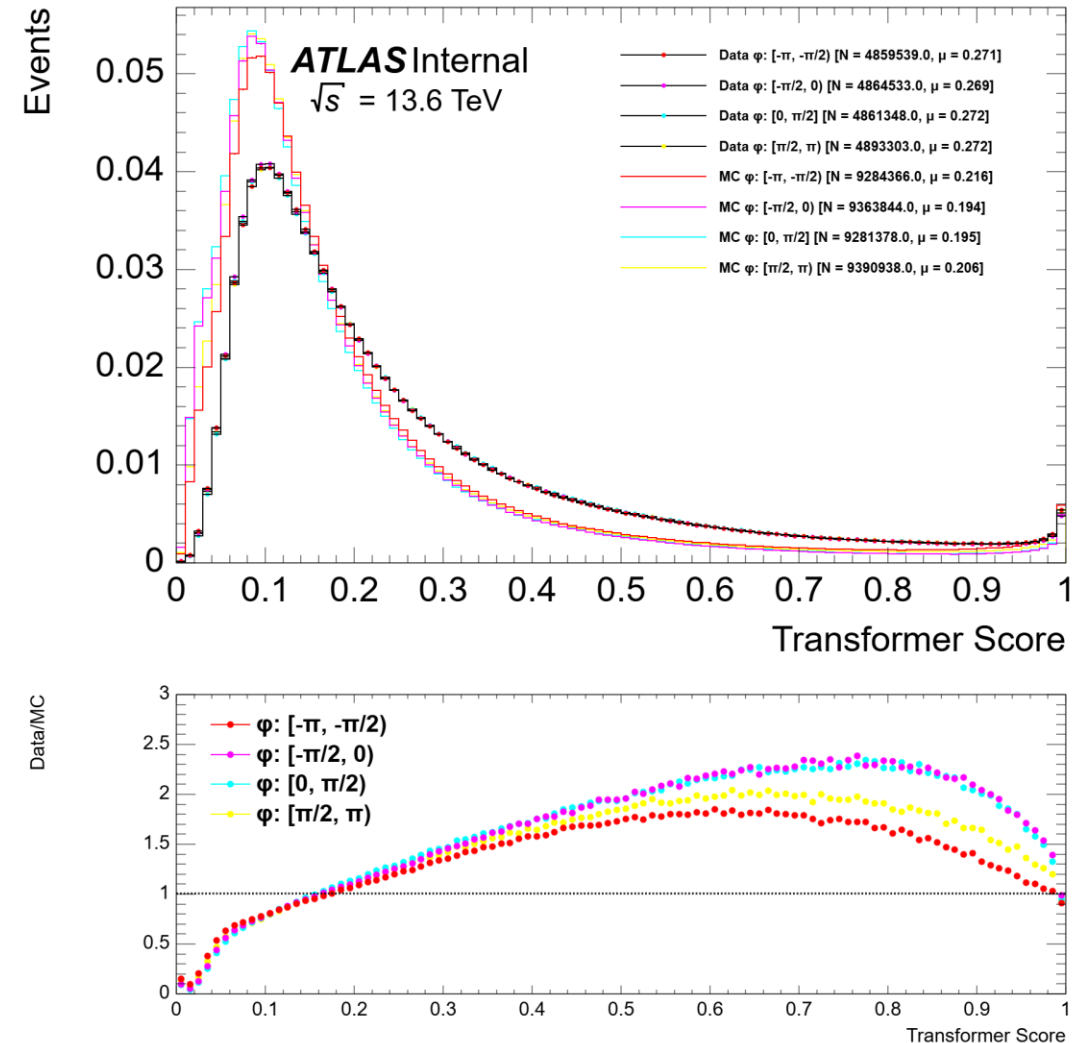
η -Binned Transformer Score

- η Bins: $[0, 0.4)$, $[0.4, 0.8)$, $[0.8, 1.37)$, $[1.52, 2.0)$, $[2.0, 2.37)$
- There is a significantly stronger agreement between data and simulation in the $[0.8, 1.37)$ bin



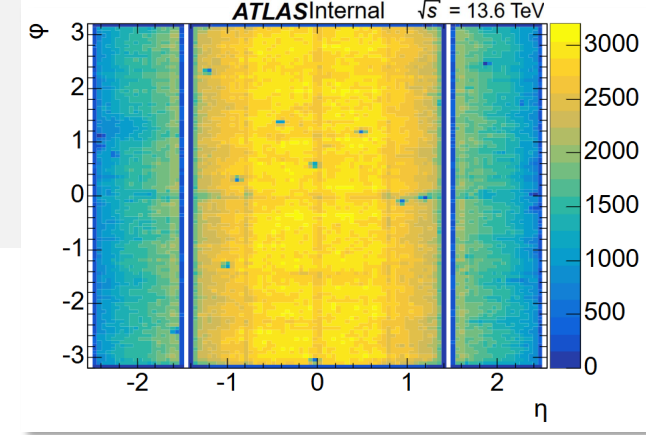
Φ -Binned Transformer Score

- Φ Bins: $[-\pi, -\pi/2)$, $[-\pi/2, 0)$, $[0, \pi/2)$, $[\pi/2, \pi)$
- Data and simulation show individual consistency across bins
- The disagreement between data and simulation remains in all bins

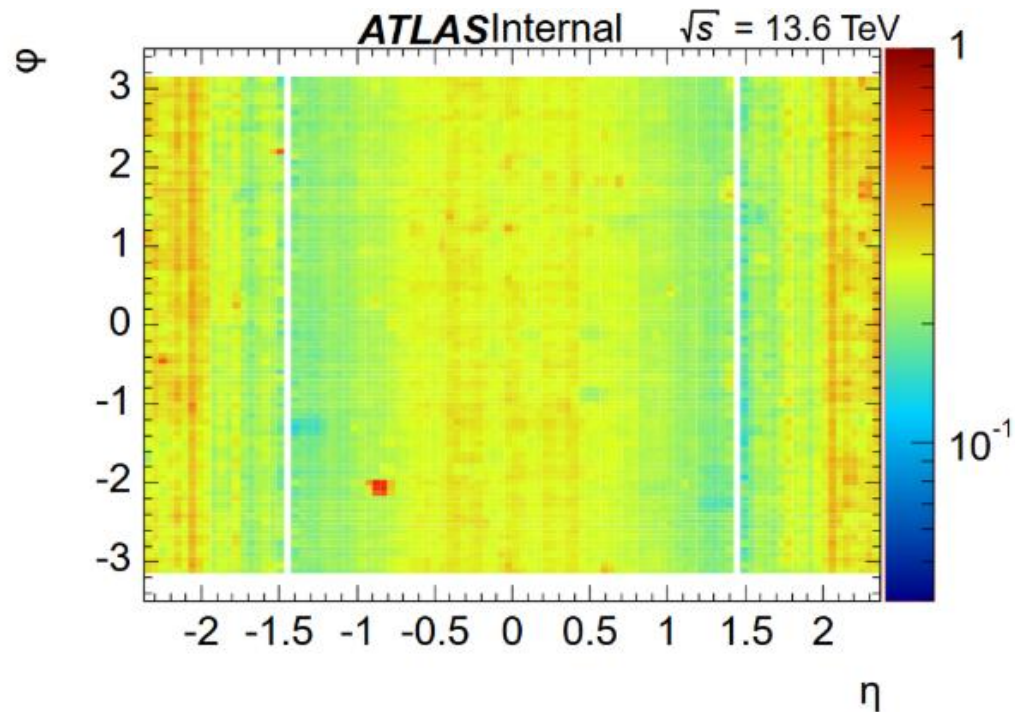


A Spatial Representation of Transformer Score

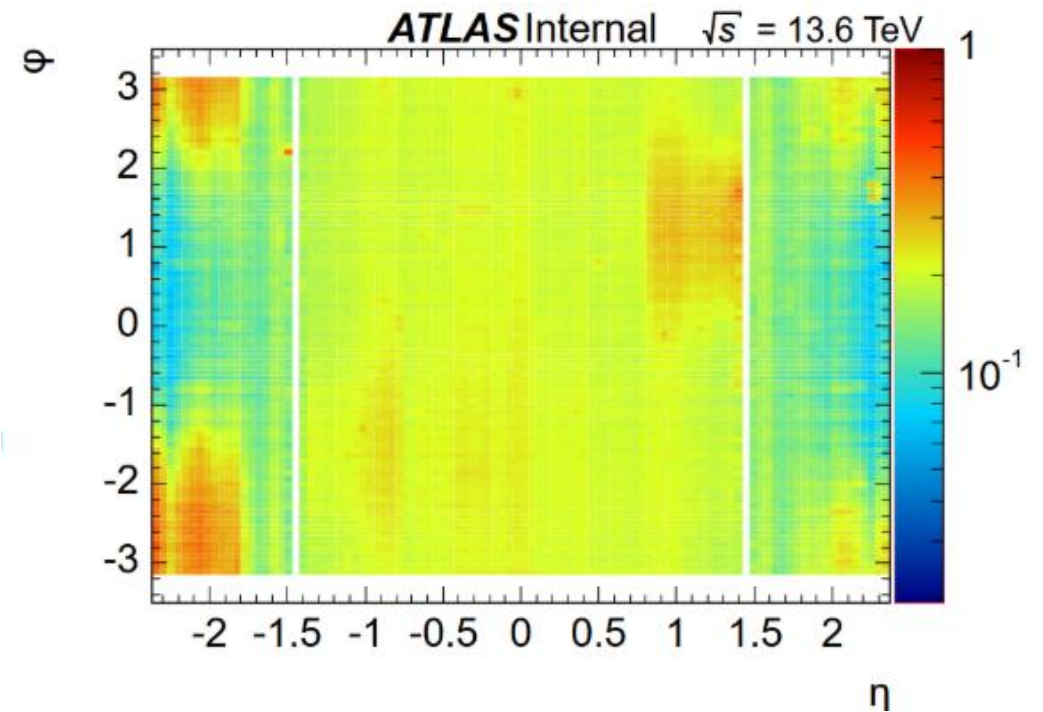
- There is spatially-correlated disagreement between data and simulation transformer scores



Data

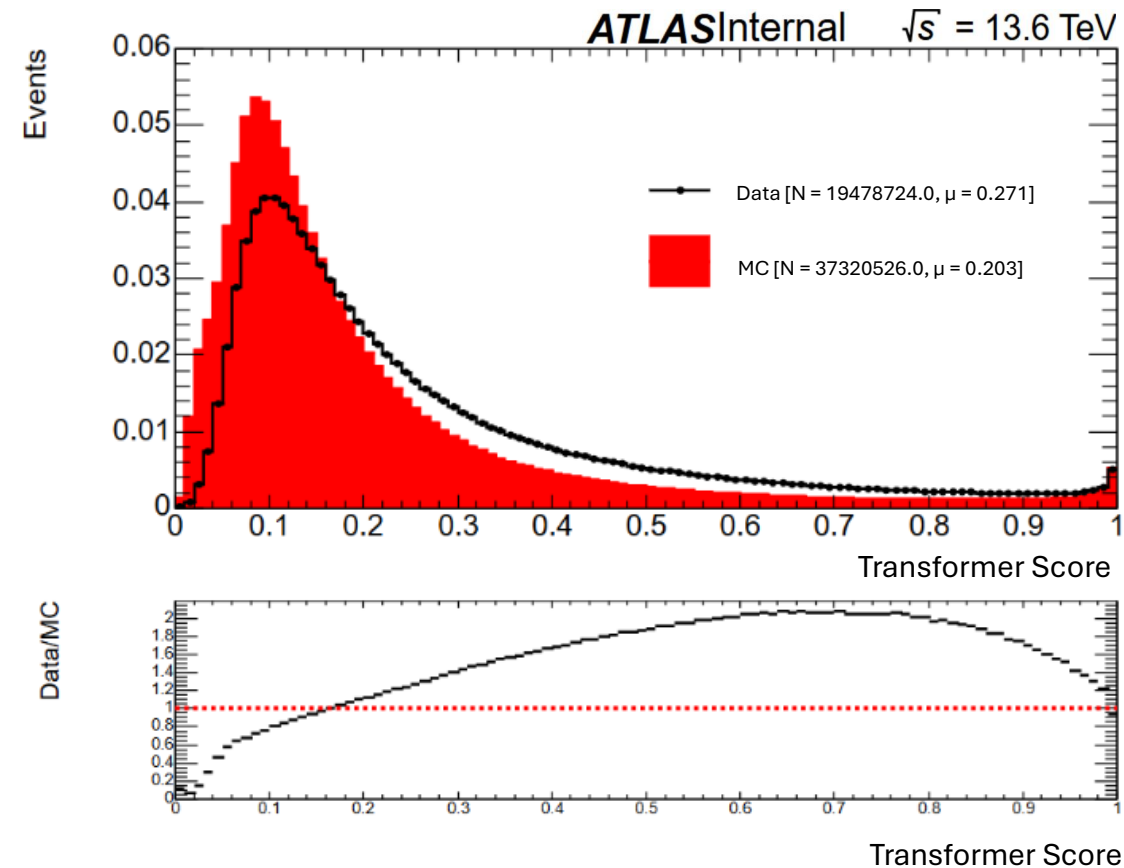


Simulation



Conclusions

Transformer Score for $Z \rightarrow e^+e^-$ Sample

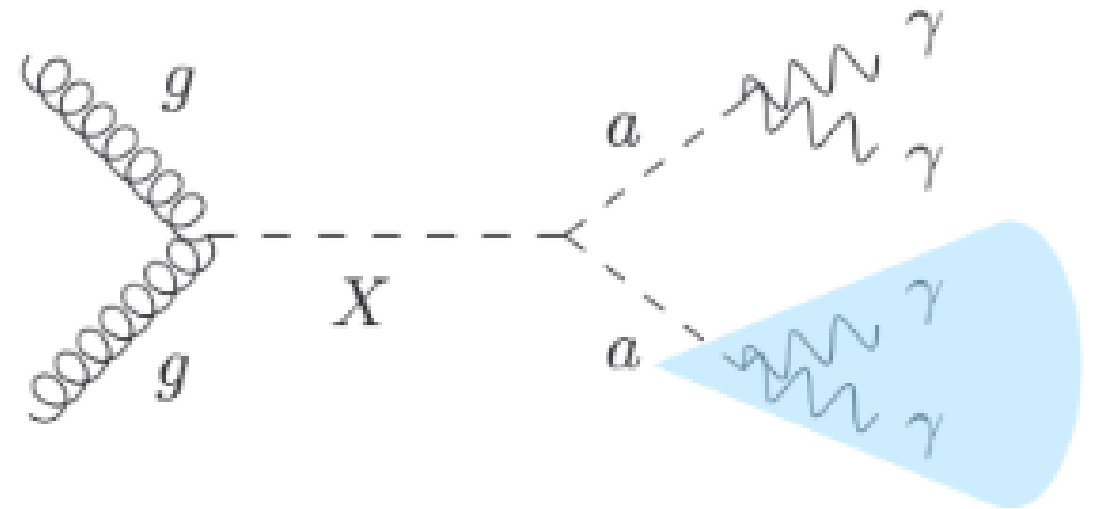


Despite good agreement between data and simulation events in kinematics and SSVs, we see:

- An unexpected magnitude of **disagreement** between data and simulation scores at high p_T
- **Spatially-localized regions** of transformer score disagreements between data and simulation

Future Work

- Re-sample along p_T and η in the training samples to match signal and background distributions
- Re-train the transformer on new samples to improve on the initial round of training
- Re-examine the improved transformer's performance on background samples



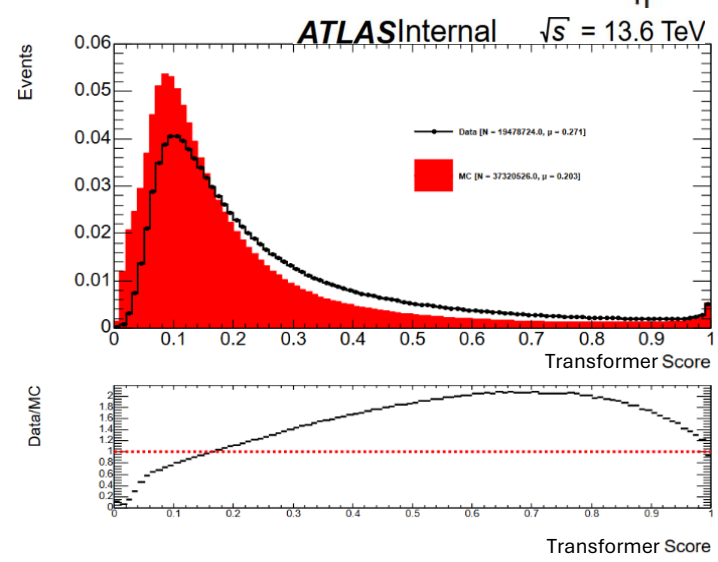
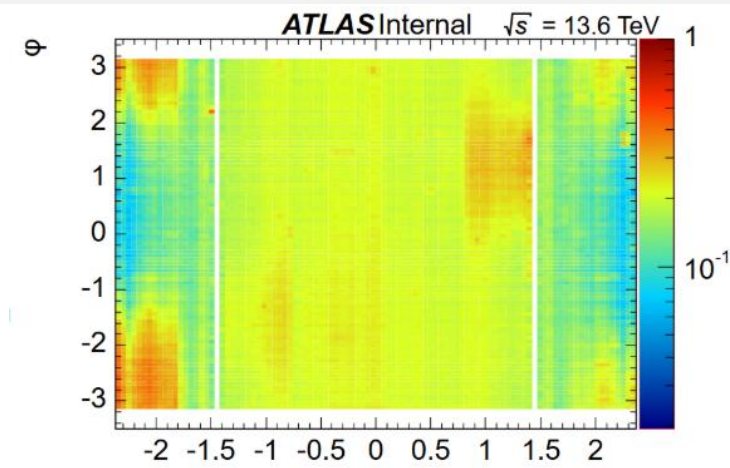
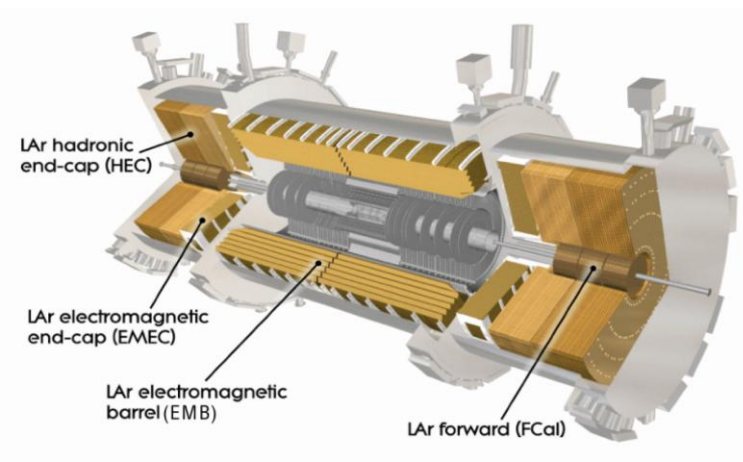
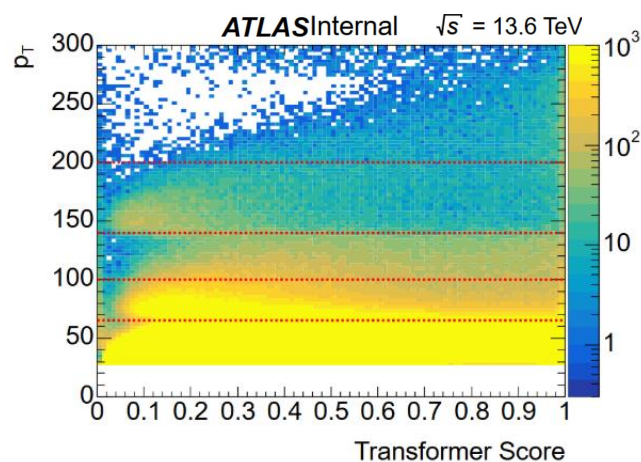
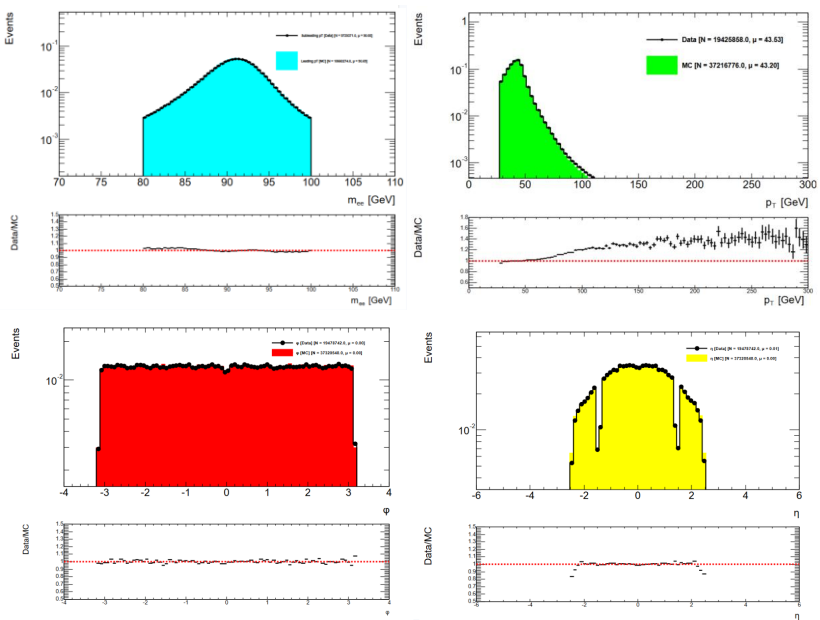
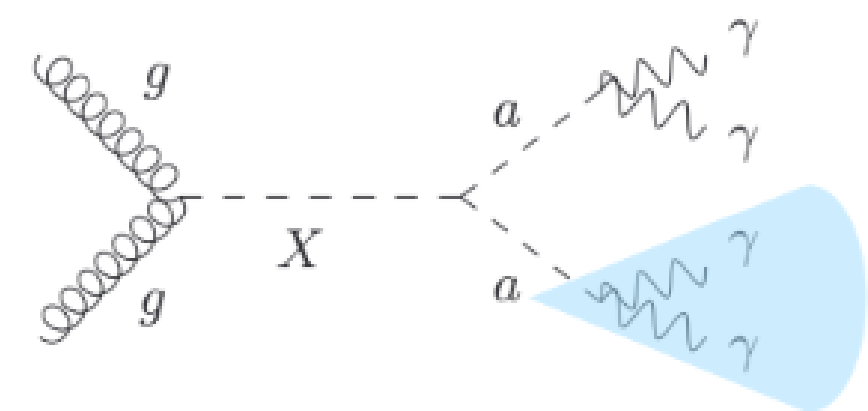
Acknowledgements

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- Professor John Parsons, Dr. Lauren Osojnak, Dr. Jonathan Long, and Gabriel Matos for amazing mentorship and support throughout this project
- Professor Georgia Karagiorgi, Professor Reshmi Mukherjee, and Amy Garwood for all of the hard work that made this REU possible
- My fellow REUs and everyone at Nevis for making this summer so much fun
- The NSF for funding this research

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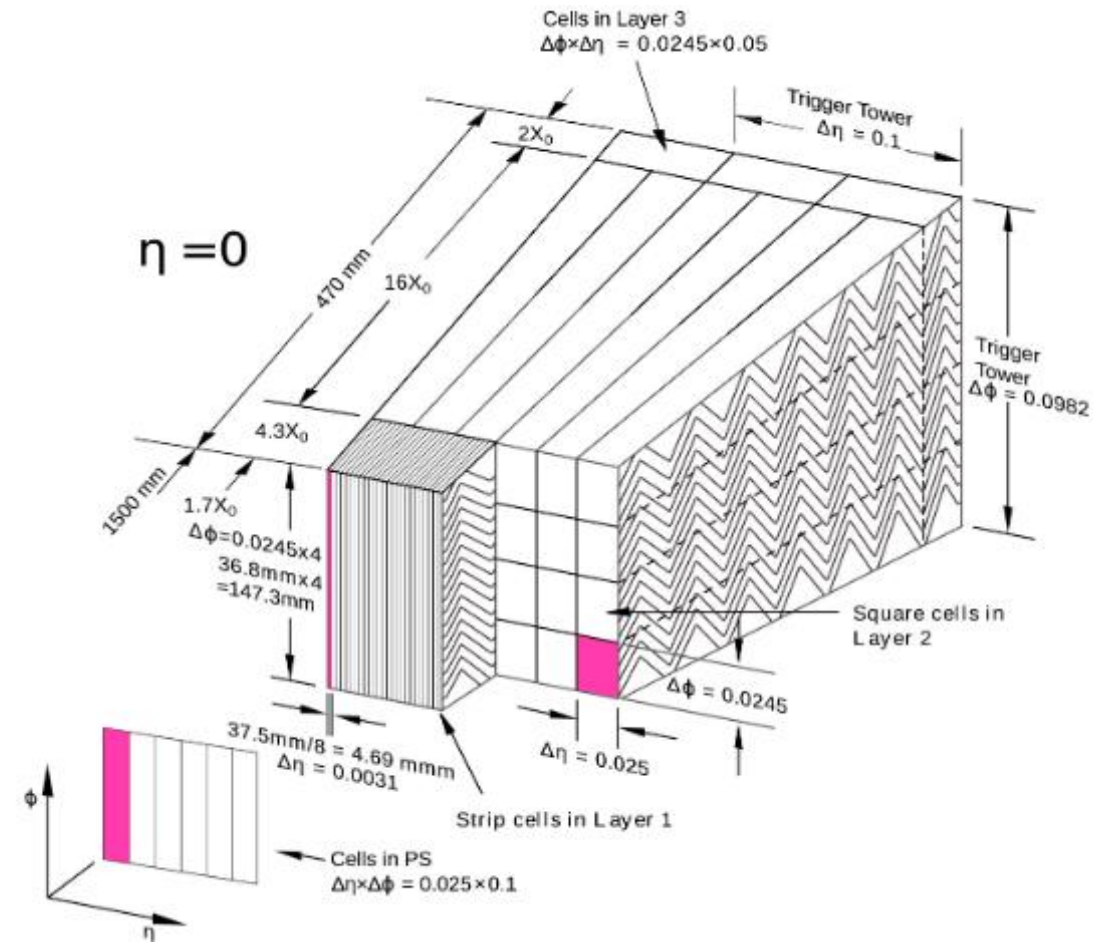
Questions?



Backup

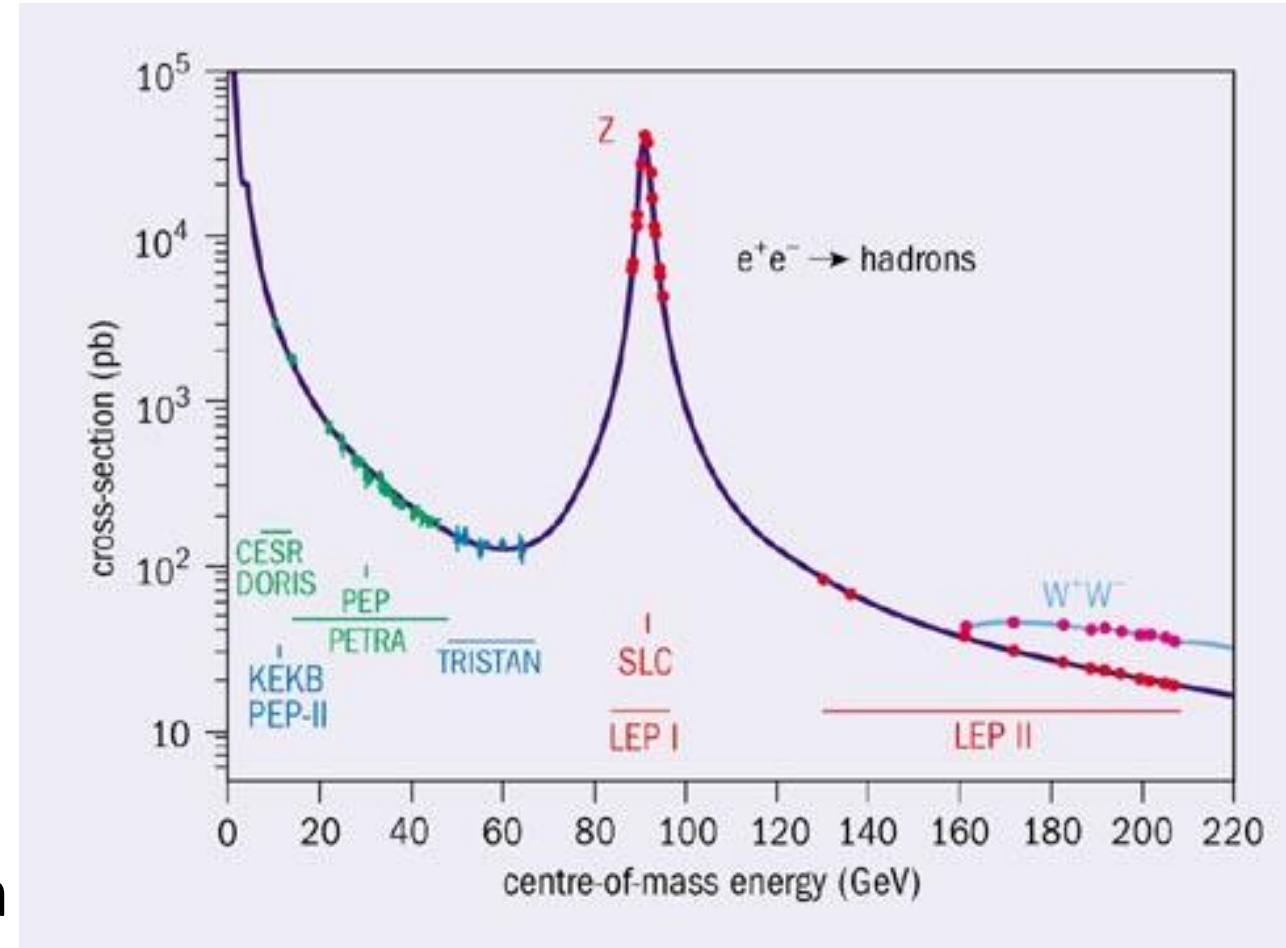
The Liquid Argon Calorimeter

- Measures energies of EM particles
- 3 Layers:
 1. Fine Granularity Strips
 2. Majority of Energy Collection
 3. High Energy Shower Correction



Z Boson Data

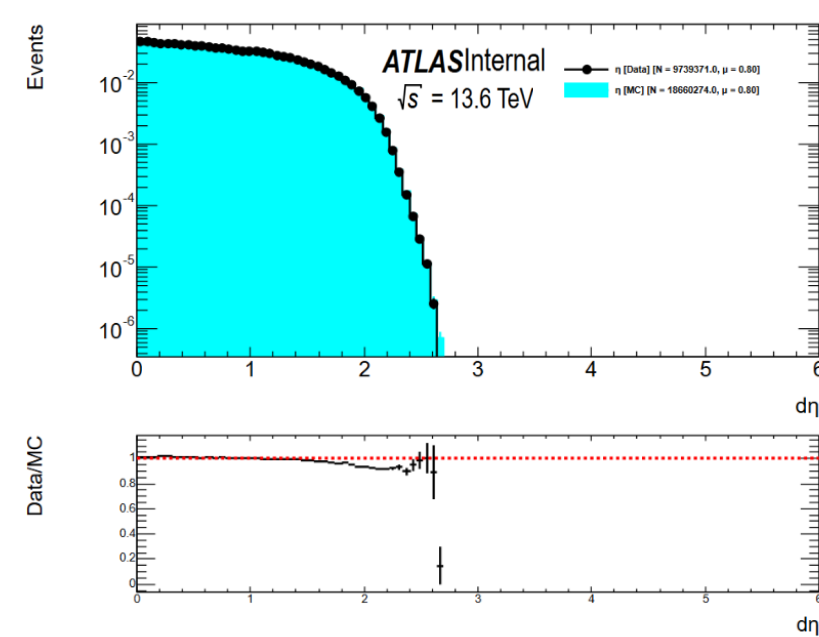
- In data, the Z boson occurs far more frequently than the noise around it
- Strongest peak: 80-100 GeV
- Allows for good assurance when using data as background validation



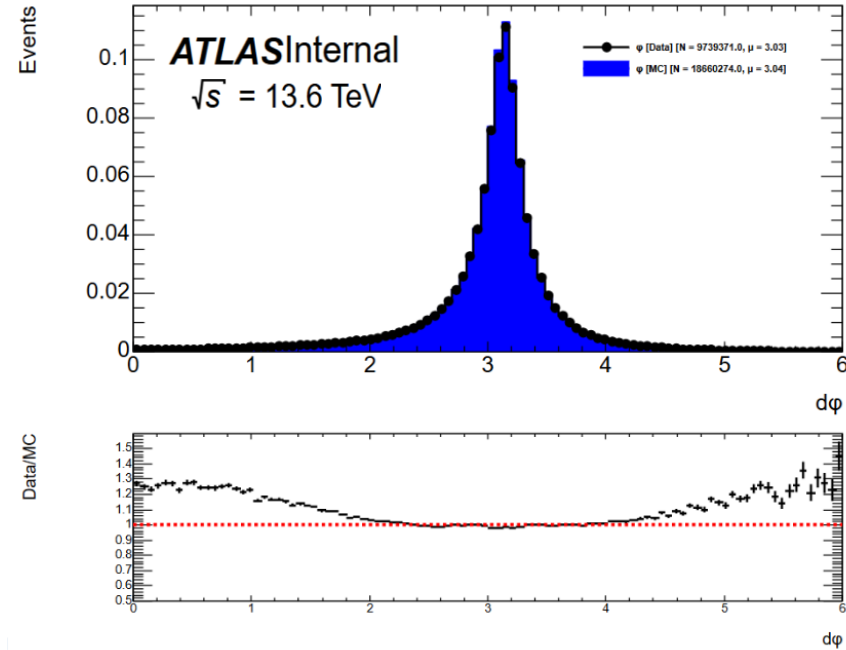
CERN Courier, 2004

Further Kinematic Plots

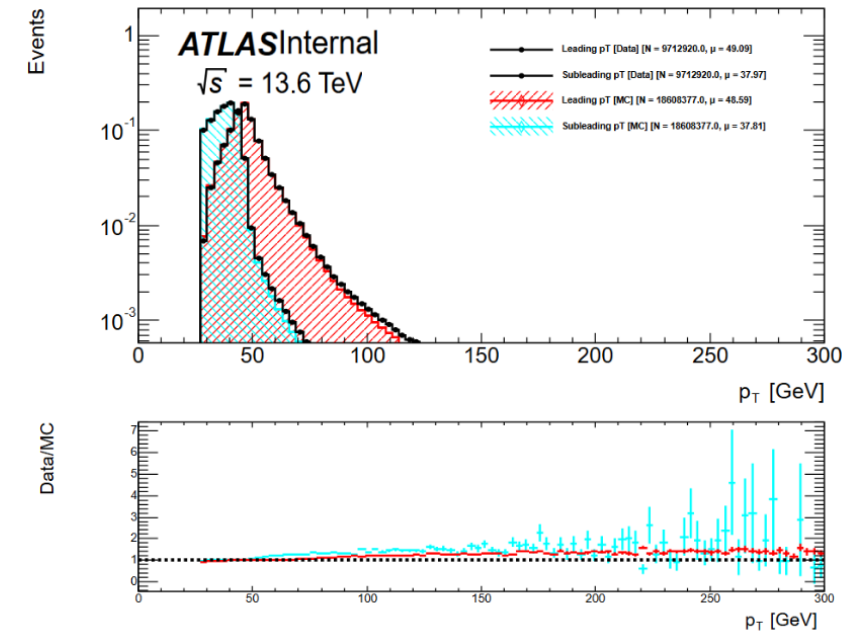
Pseudorapidity e^- Separation ($d\eta$)



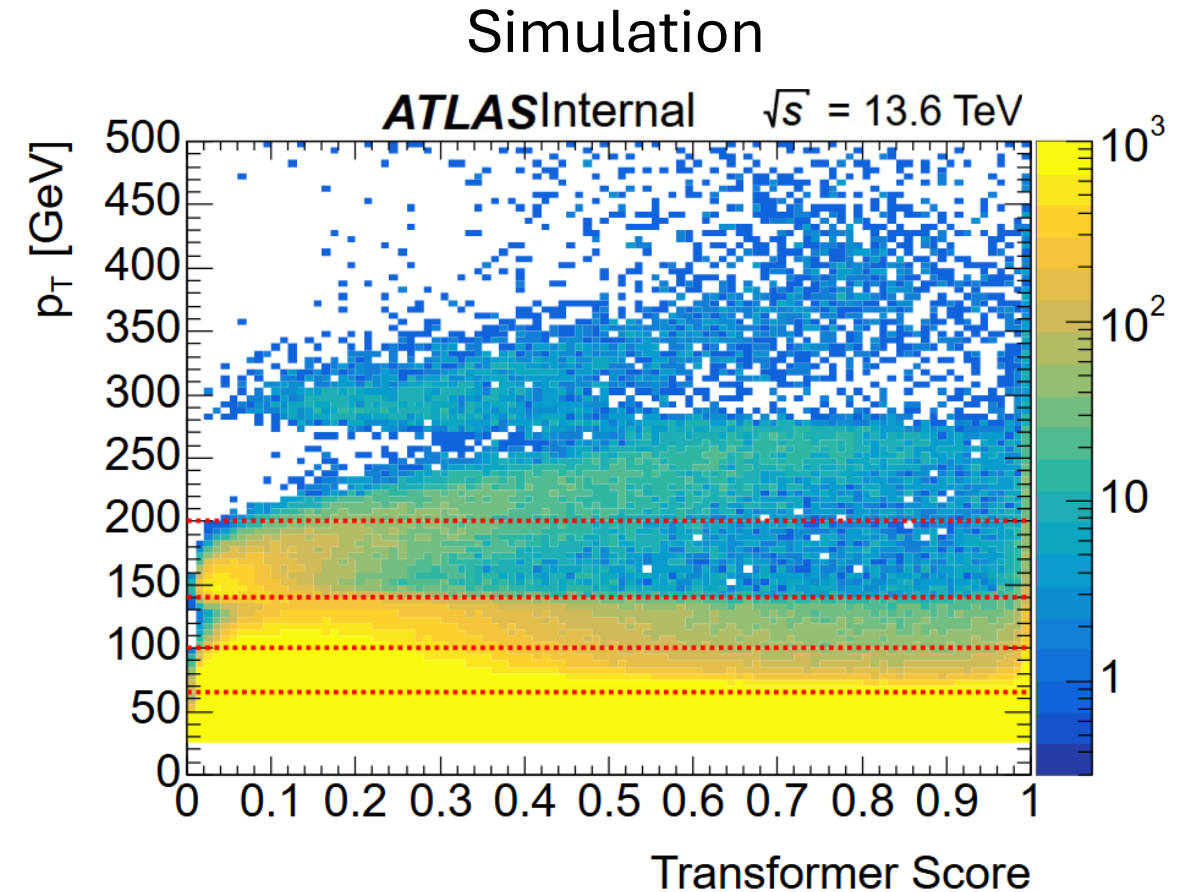
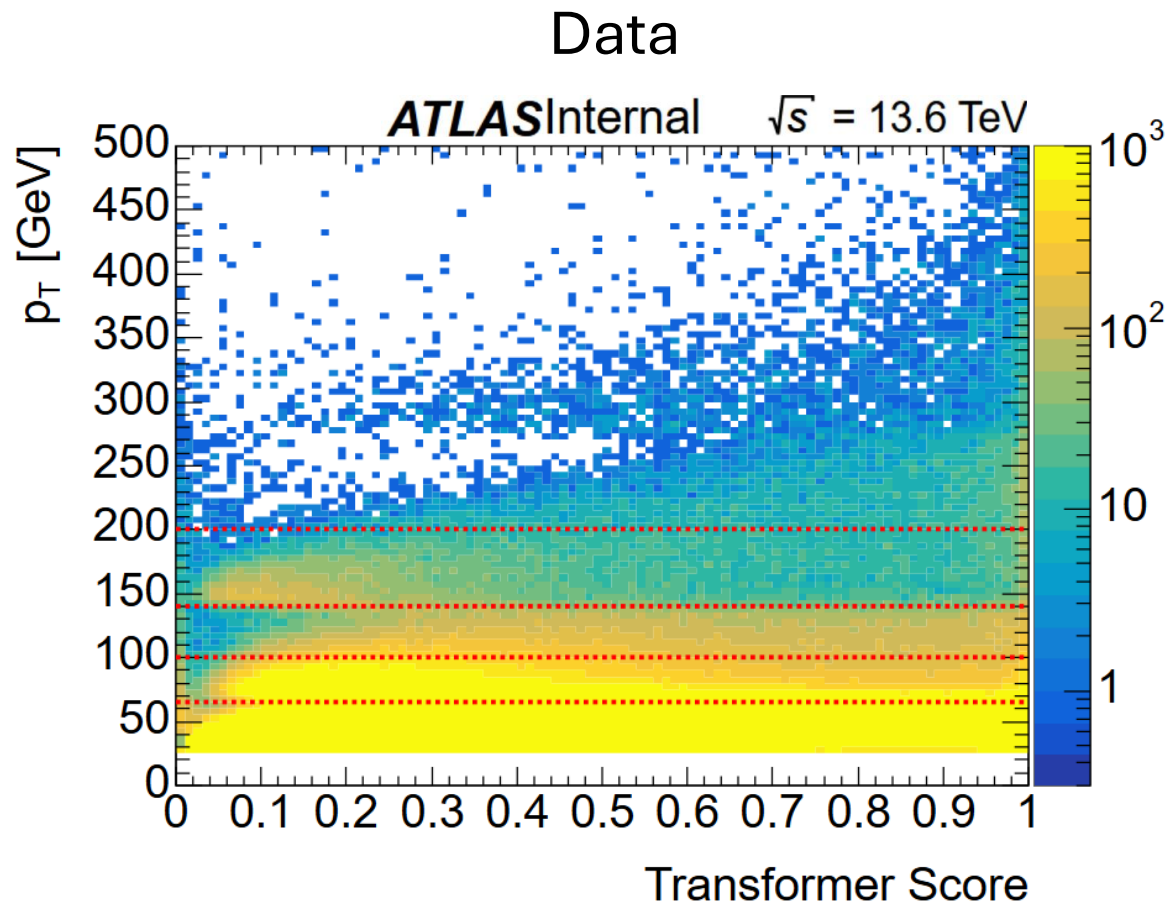
Azimuthal e^- Separation ($d\Phi$)



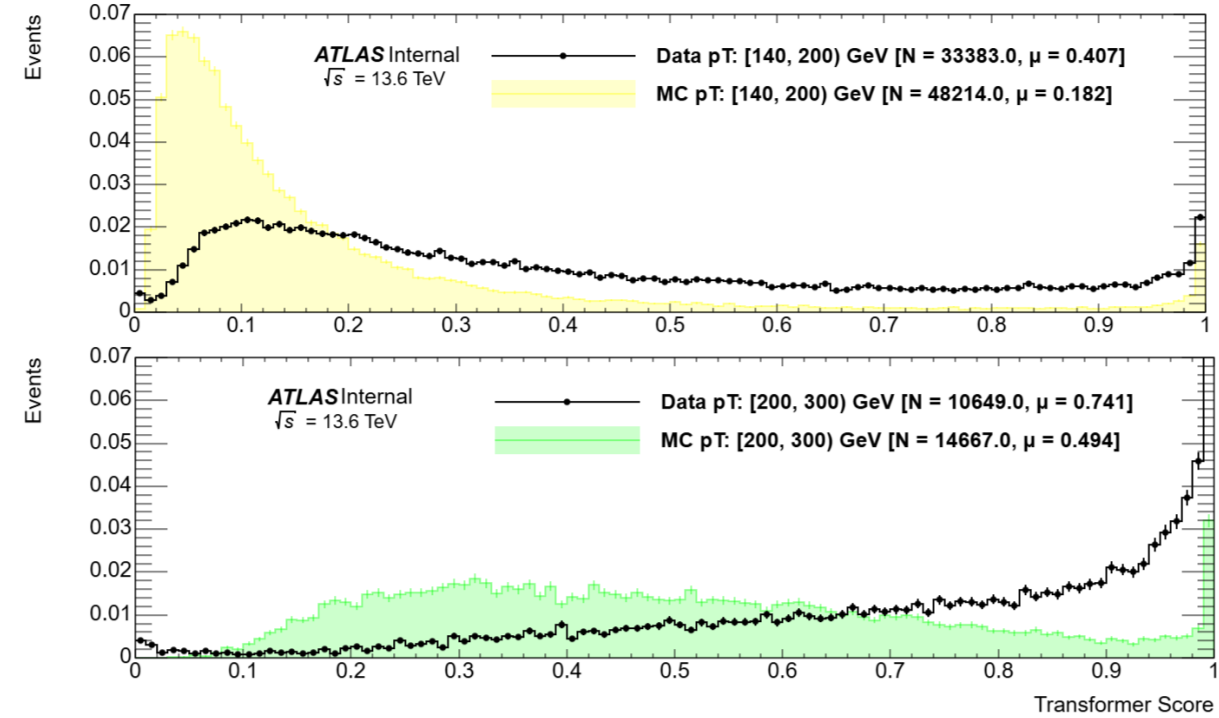
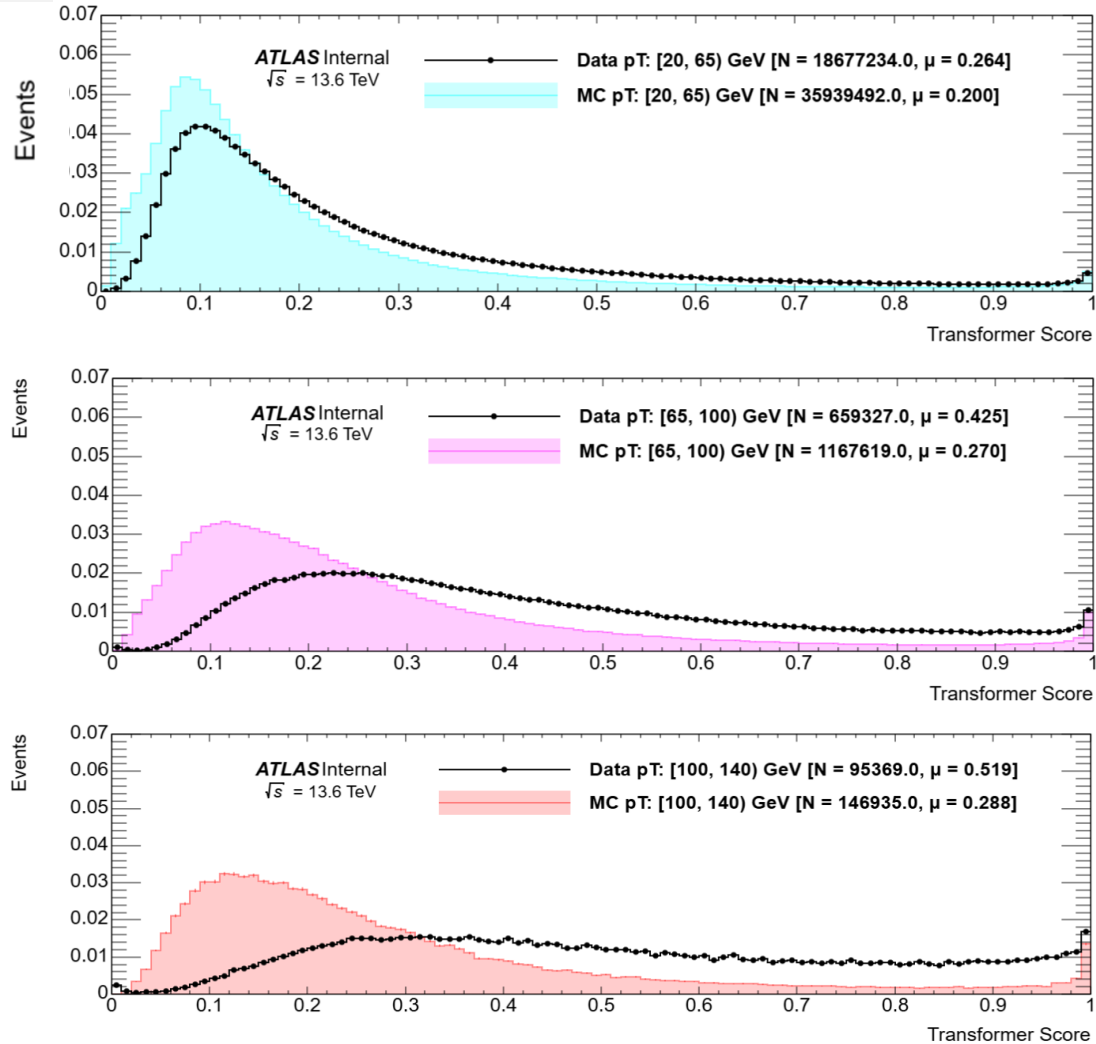
Leading and Sub-leading p_T



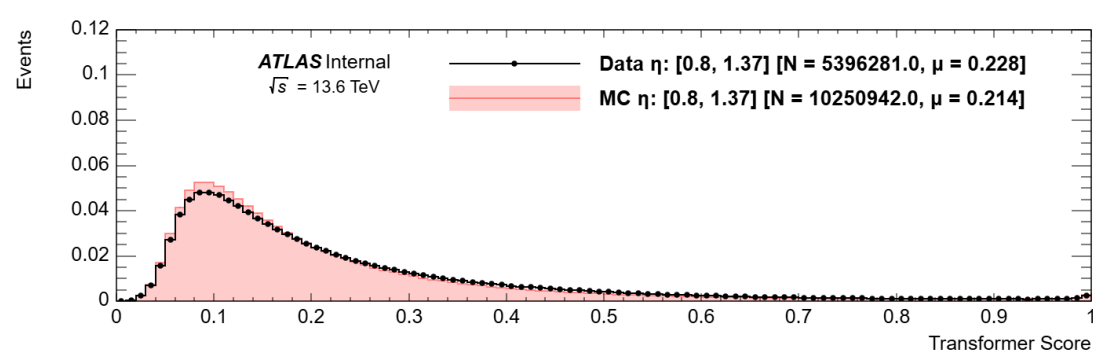
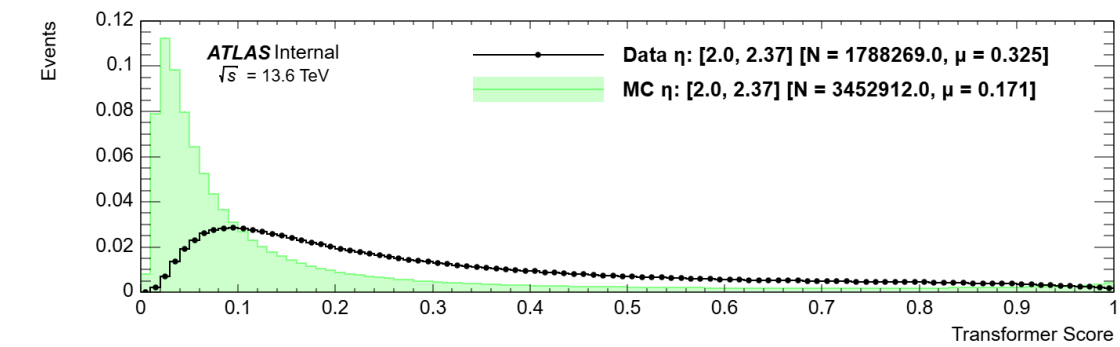
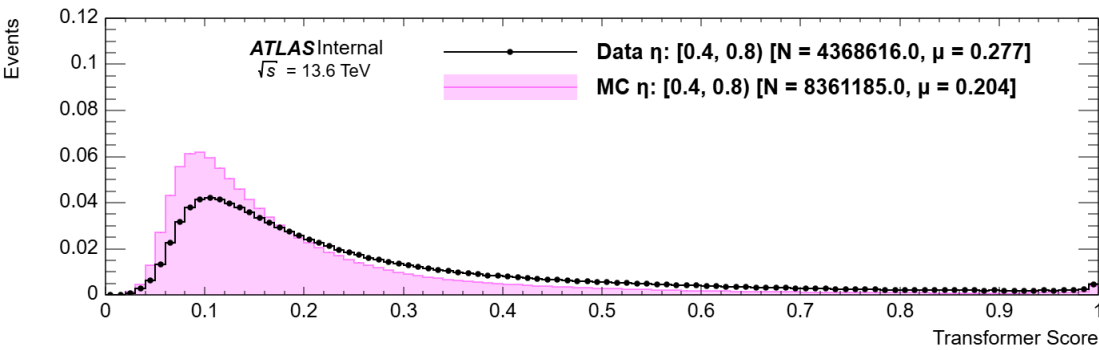
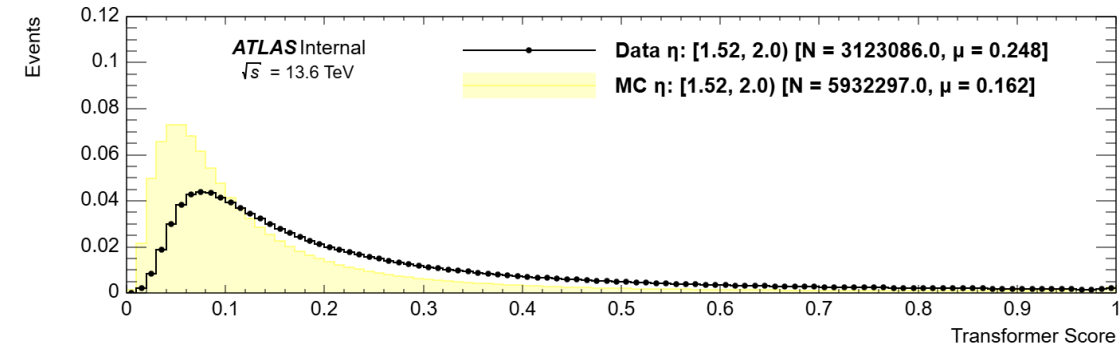
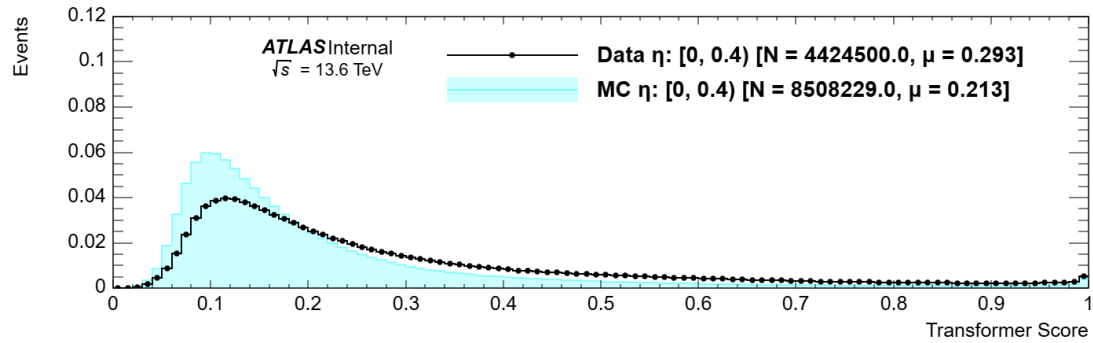
Extended p_T -Score Distribution



Transformer Score - Stratified p_T Binning

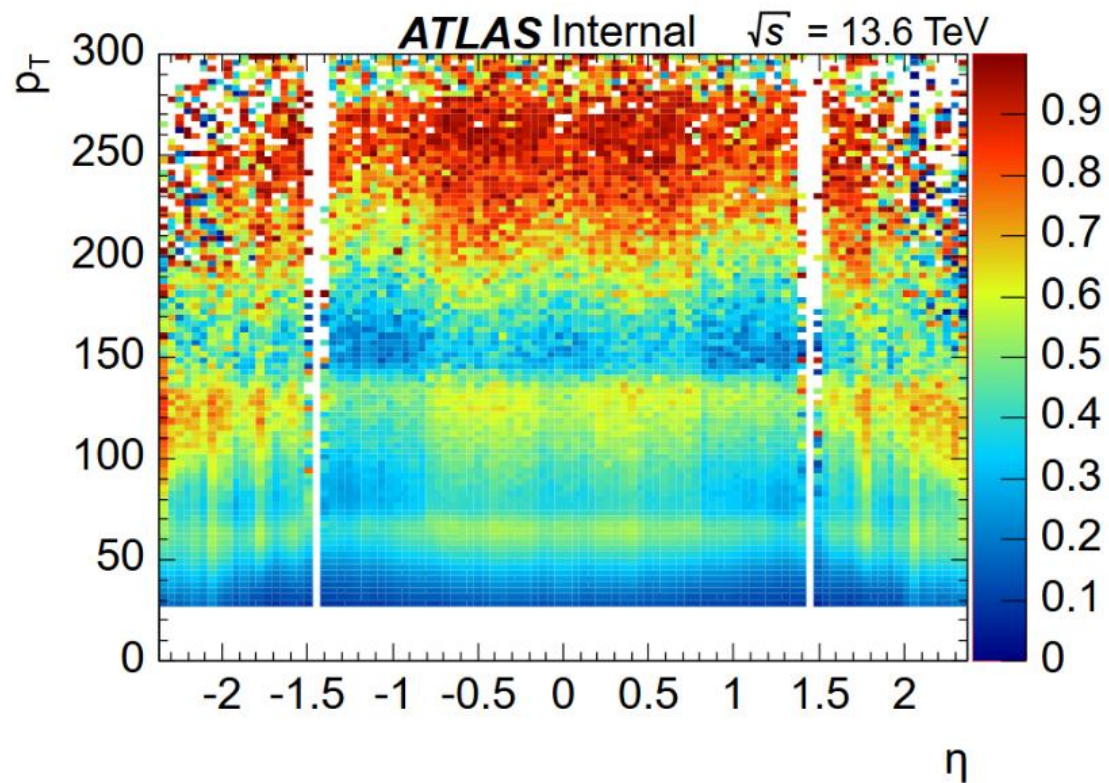


Transformer Score - Stratified η Binning

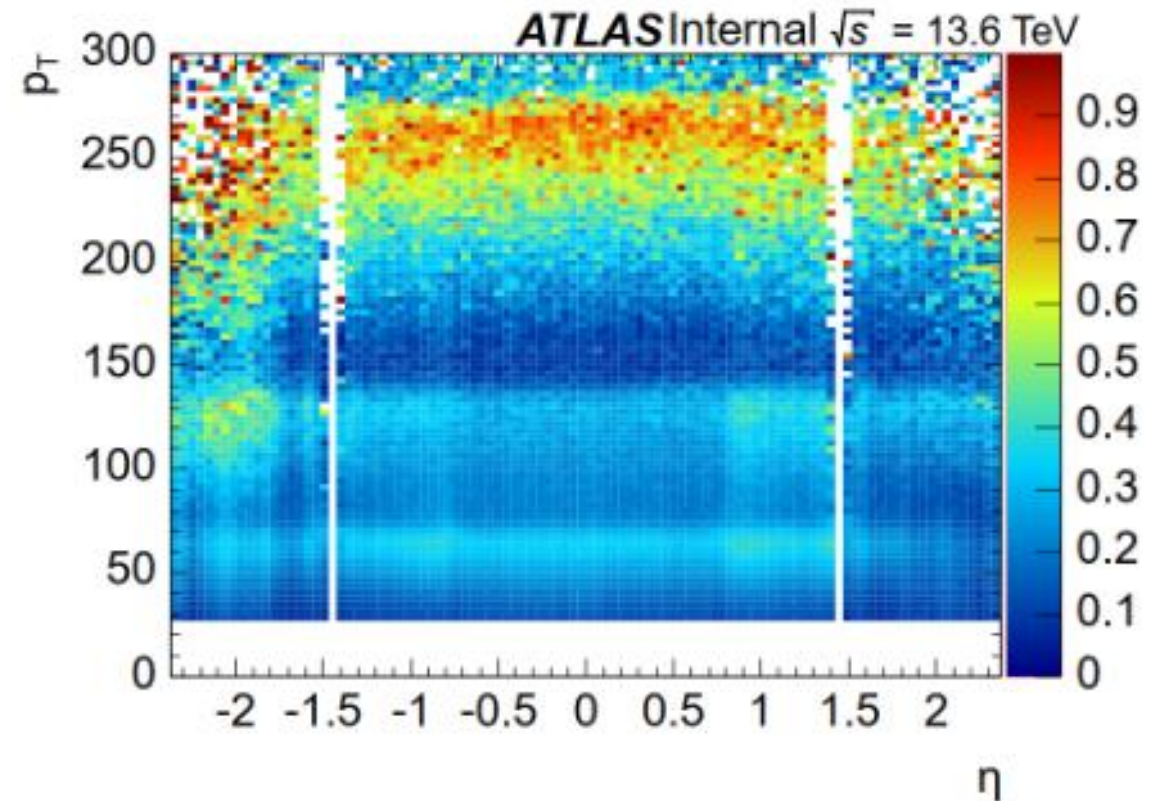


2-Variable Kinematic Score Correlations – p_T vs η

Data

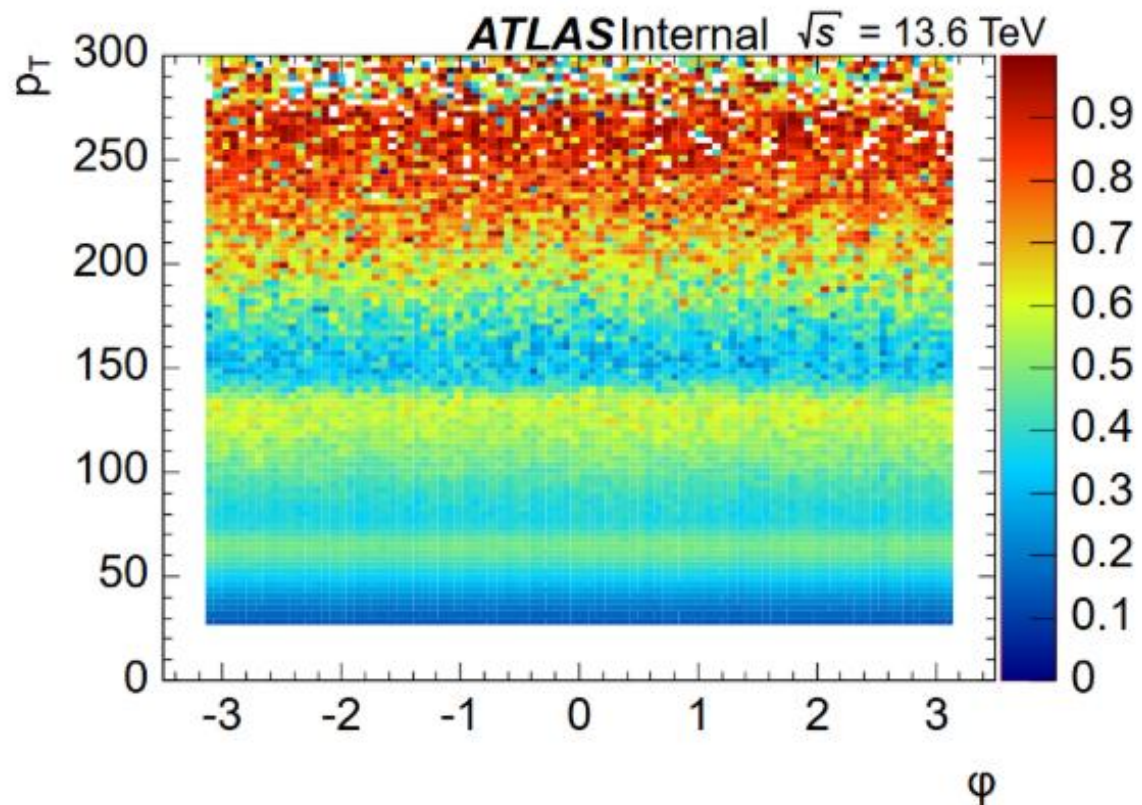


Simulation



2-Variable Kinematic Score Correlations – p_T vs Φ

Data



Simulation

