



AI-Assisted Analysis to Enhance Discovery Potential in High Energy Physics

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Motivation

Traditional LHC searches follow a supervised script: test a known signal against background. This model-dependent approach scored big wins, like the Higgs, but has yet to uncover new physics.

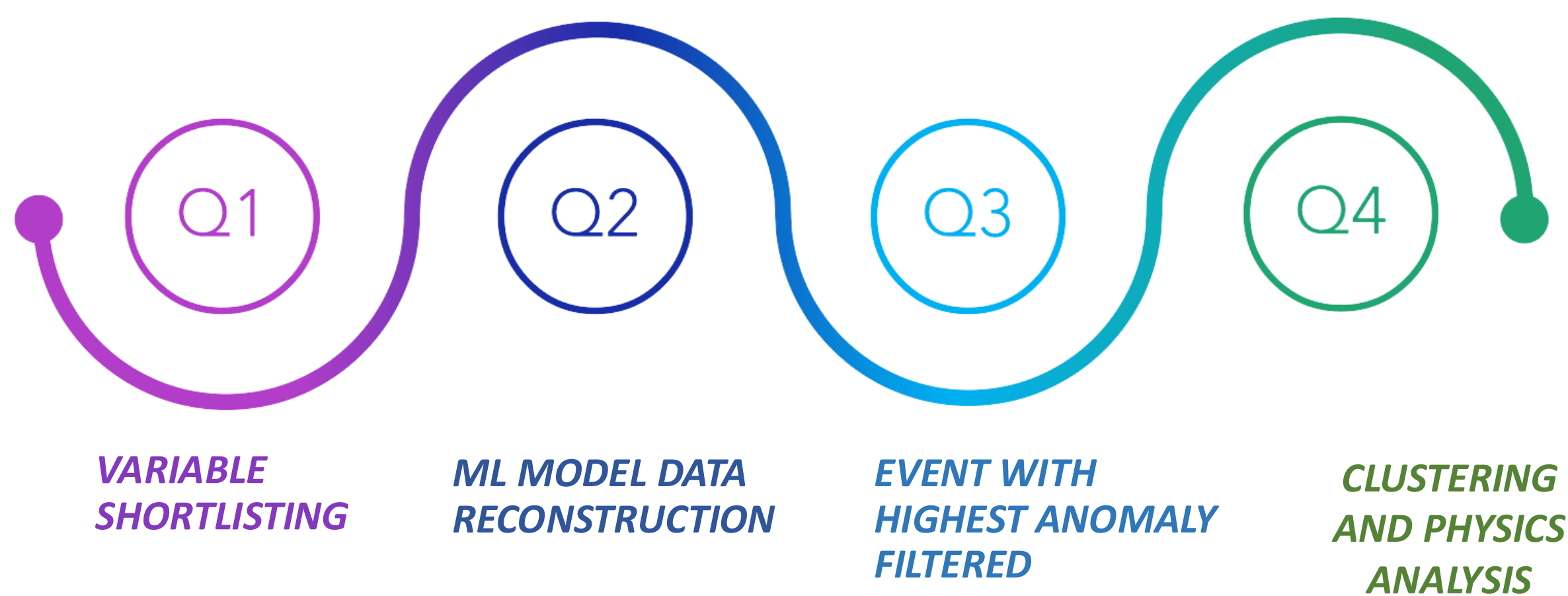
The challenge? The BSM landscape is vast, and fixed triggers or cuts can bias what gets seen. A Novel approach: **unsupervised, model-agnostic strategies**. Powered by **generative deep learning**, these methods learn the patterns from data and flag the outliers. We aim to catch **the rare, the BSM event**.

Novelty

We present an **unsupervised, transformer-based** reconstruction model trained solely on Run 2 2015 proton-proton collision data from the ATLAS experiment.

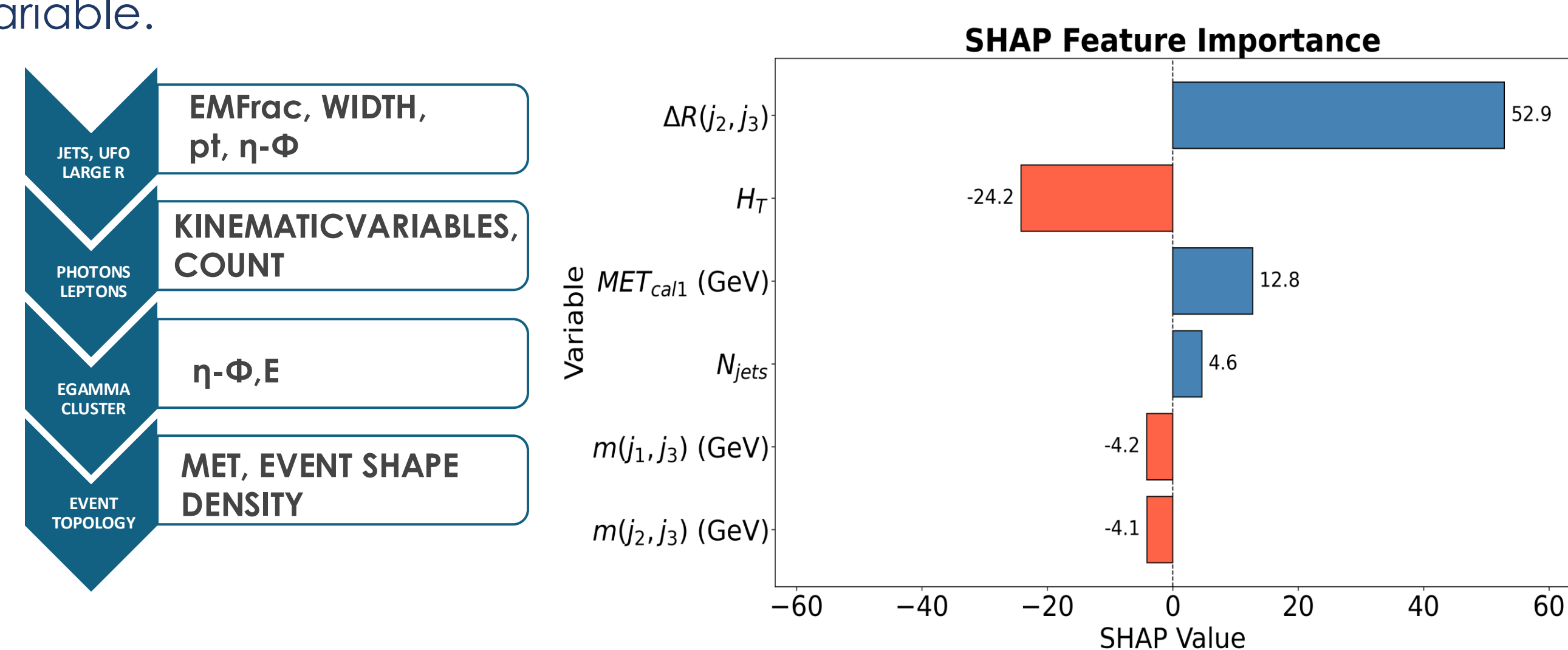
We developed a **Masked Tab-Transformer [1]** that handles **high-dimensional, sparse** tabular detector data, aiming to reconstruct events with large fractions of **missing inputs**. Our model reconstructs all detector-level features: jets, leptons, photons, missing energy, calorimeter deposits, and event topology.

Model Development Workflow



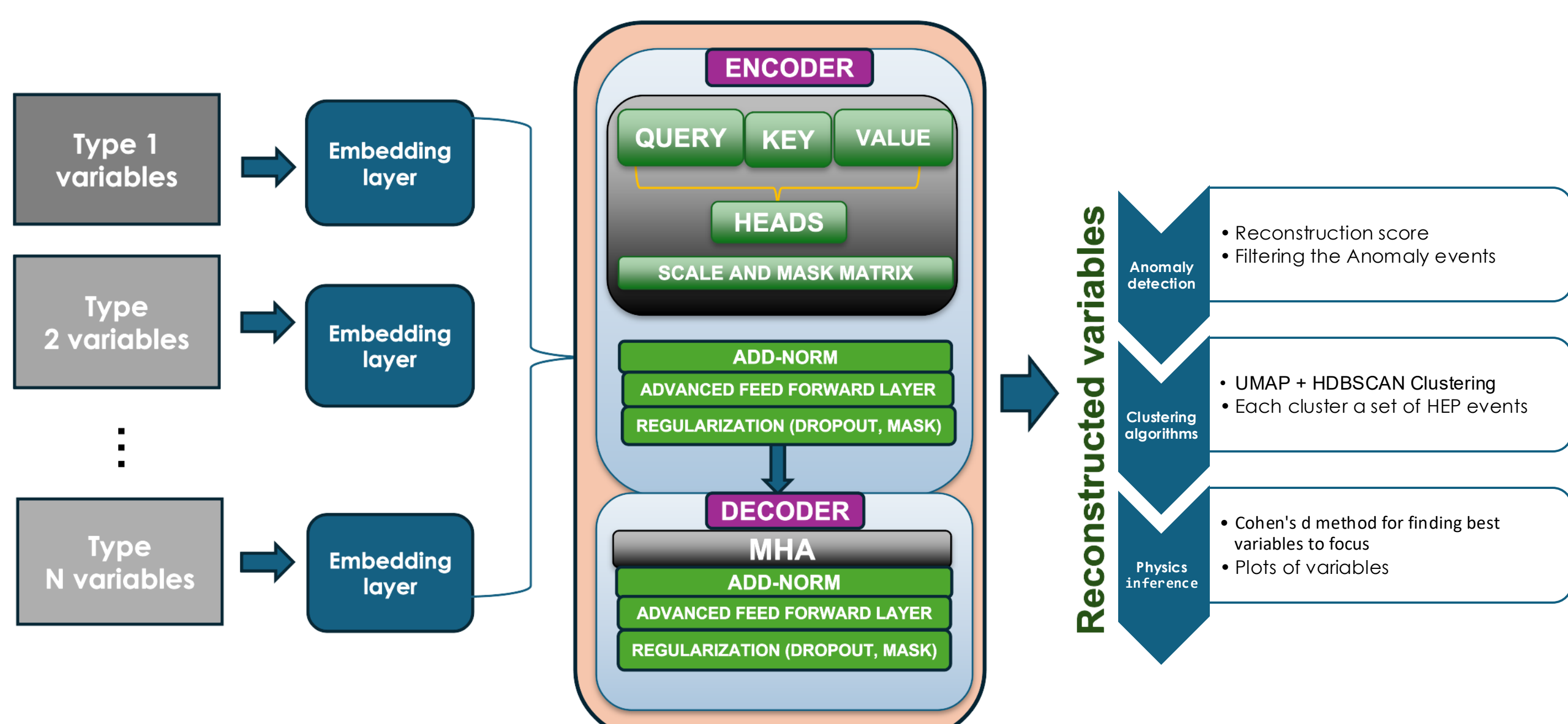
We analyze the **ATLAS Run-2 open dataset** (13 TeV, 2015) [2] with ~5 million proton-proton collisions and full detector readouts. From each event, we extract low-level feature jets, leptons, photons, MET, calorimeter deposits, and event shape density.

- Preprocessing and scaling using **ROOT RDataFrame** on **CERN's SWAN** interfaced with **Dask** and **HTCondor**.
- Streaming output **.root** files directly as TensorFlow via **TMVA TF batch generator** [5].
- SHAP**-based feature selection to reduce redundancy and remove noisy variable.



Tab-Transformer autoencoder with masked self-attention and custom masked loss (weighted loss).

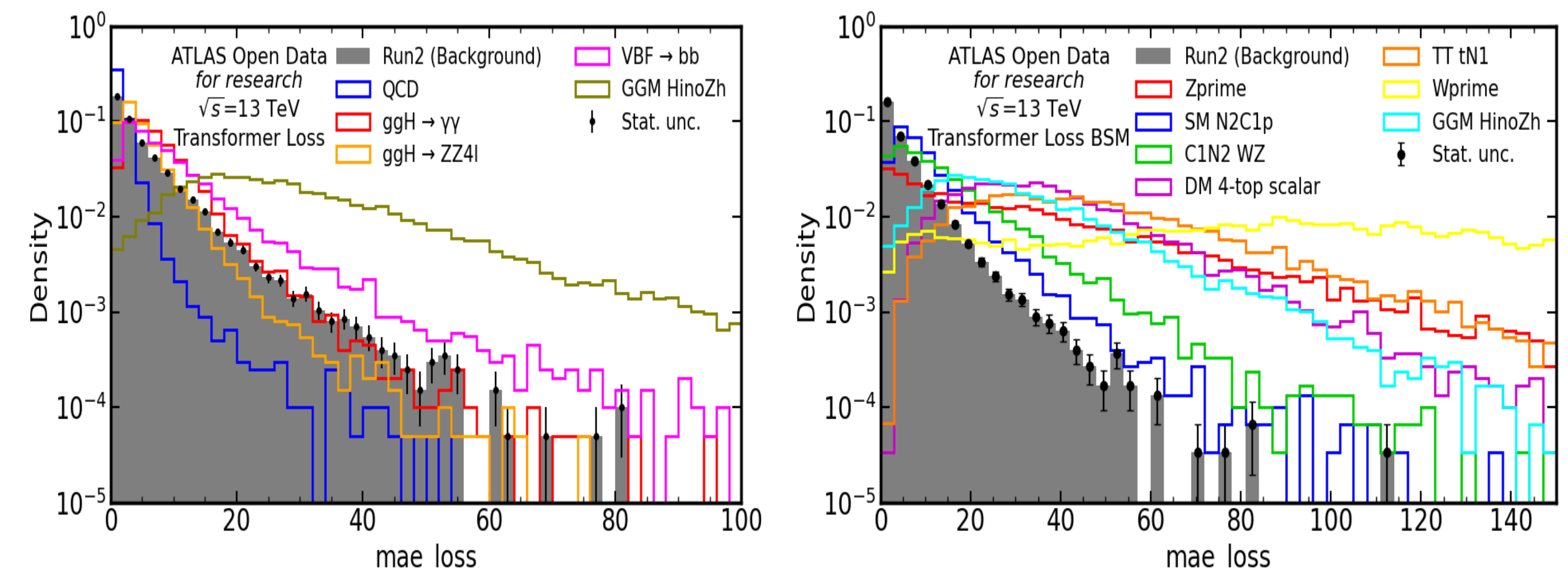
- Outliers are organized with **UMAP[3] + HDBSCAN[4]**, revealing structured patterns (e.g., high- p_T leptons, jet-rich events).
- Cohen's d** quantifies statistical significance, highlighting clusters that truly deviate from SM pattern



Results

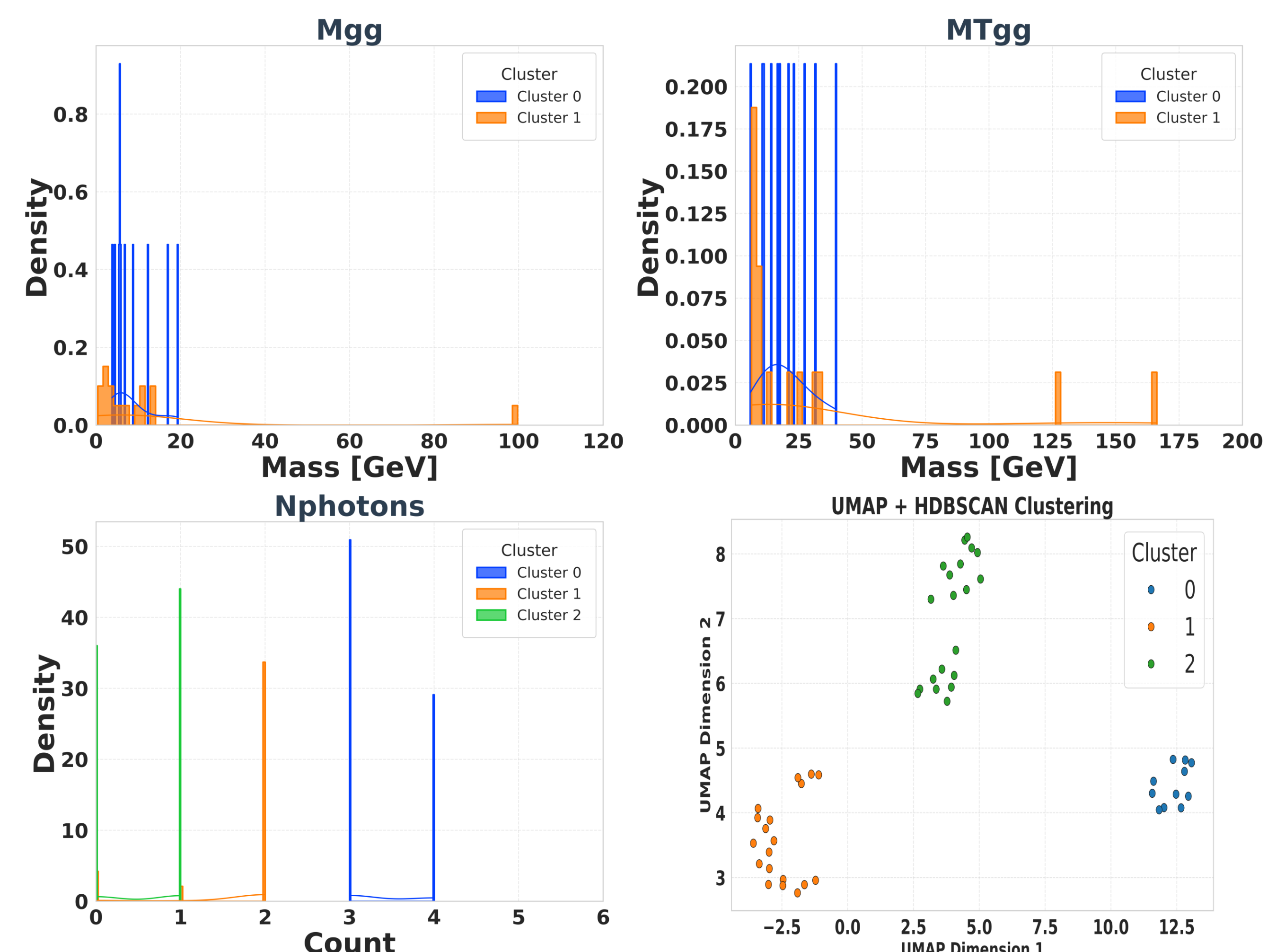
Testing on simulated BSM samples (SUSY, exotic resonances, rare Higgs/top decays) treated as anomalies.

- SM backgrounds** reconstructed with high fidelity (low loss).
- BSM events** show systematically higher reconstruction loss.
- Examples: leptonic SUSY cascades and heavy resonance decays produce clear high-loss tails, well-separated from SM validation samples.



- Performed clustering analysis to identify the most important variables driving the separation.
- Based on the Cohen's d value plotted important variable in clusters.

Distributions by Cluster



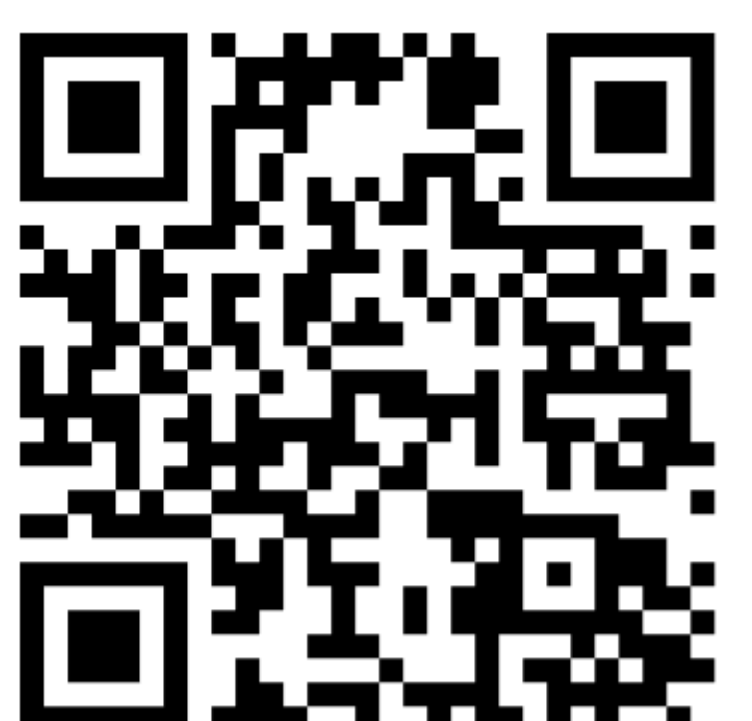
Unbiased Search

Fully unsupervised pipeline trained on raw detector-level data, with no assumptions on final states.

- Transformer autoencoder** learns the Standard Model features from training events, flags outliers via reconstruction loss acting as a data-driven trigger for rare or unknown phenomena.
- UMAP + HDBSCAN** cluster anomalies, revealing structured event classes.
- Scalable & interpretable**, expanding discovery reach in underexplored phase space corners.
- Future: larger datasets**, Improve network architecture, calibration, **richer inputs (tracking, timing)**, **real-time trigger-level deployment** for truly model-independent new physics searches

Reference

- [1] Arik & Pfister, *TabTransformer*, AAAI 2021, arXiv:2012.06678.
- [2] ATLAS Open Data, *pp collisions at $\sqrt{s} = 13$ TeV, Run 2* (2015–2016), doi:10.7483/OPENDATA.ATLAS.9HK7.P5SI
- [3] L. McInnes et al., arXiv:1802.03426 [stat.ML]
- [4] R.J.G.B. Campello et al., *Adv. Knowl. Discov. Data Min.* (2013), Springer, ISBN 978-3-642-37456-2
- [5] ROOT TMVA, RBatchGenerator TF (2024), root.cern.ch



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