

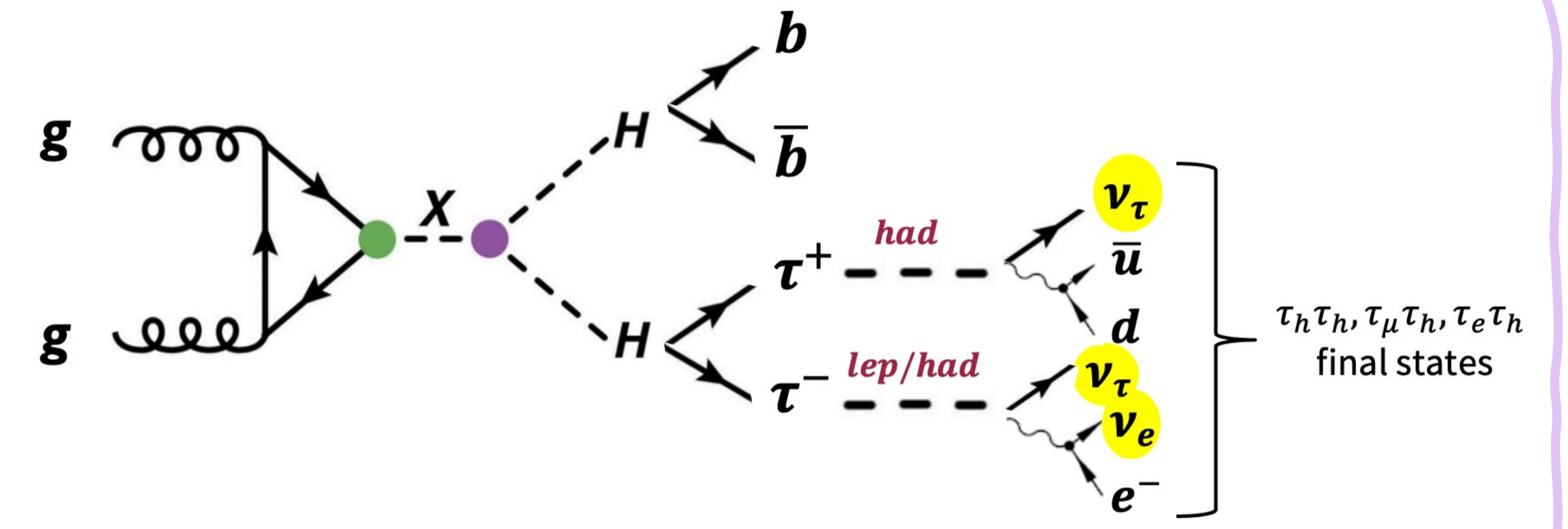
Particle Transformer for τ lepton pair invariant mass reconstruction for the $HH \rightarrow b\bar{b}\tau^+\tau^-$ CMS analysis

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Introduction

- One of the most interesting channels to probe theories beyond the Standard Model at LHC, is the production of a new massive particle, that decays into pairs of Higgs Bosons (HH) which, in turn, decay into a pair of b -quarks and a pair of τ leptons ([1])
- A fundamental discriminant variable to separate HH signal from the backgrounds is the invariant mass of the di- τ system ($m_{\tau\tau}$), which can be reconstructed starting from the decay products of each τ lepton. However, the presence of neutrinos in the final state determines a lack in terms of energy which leads to an underestimation of the invariant mass itself
- The current used algorithm to reconstruct the $m_{\tau\tau}$ variable in CMS is SVFit^[2] that, based on a maximum likelihood approach, predicts the most probable kinematics of the neutrinos taking as inputs the visible decay products of each τ lepton and the Missing Transverse Momentum (MET)
- The relevant computation time combined with a reduced resolution of the reconstructed mass by SVFit opens opportunities for new **ML-based strategies**



SIGNAL: BSM Resonant HH production via gluon-gluon fusion (ggF)

$$M_X \in [250 \text{ GeV} - 550 \text{ GeV}]$$

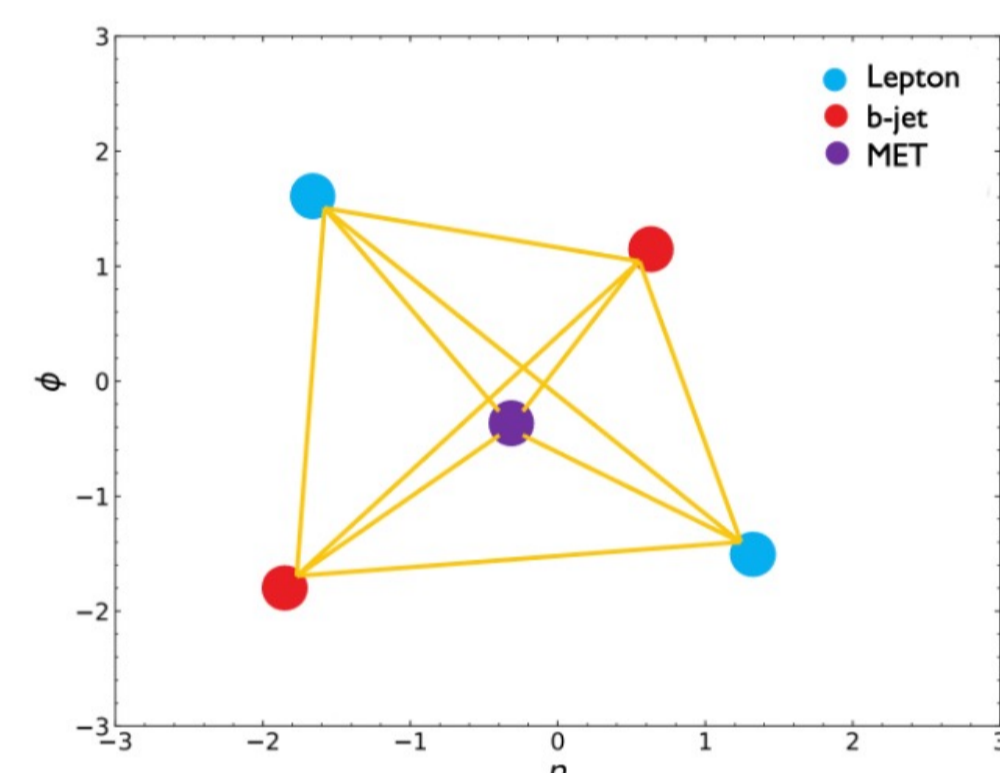
BACKGROUND: Drell-Yan, $t\bar{t}$ events

This poster propose a new technique: a deep learning model – Particle Transformer (ParT) – to estimate the four-momenta of the neutrinos involved in the decay of the di- τ system for a high-resolution reconstruction of the corresponding invariant mass

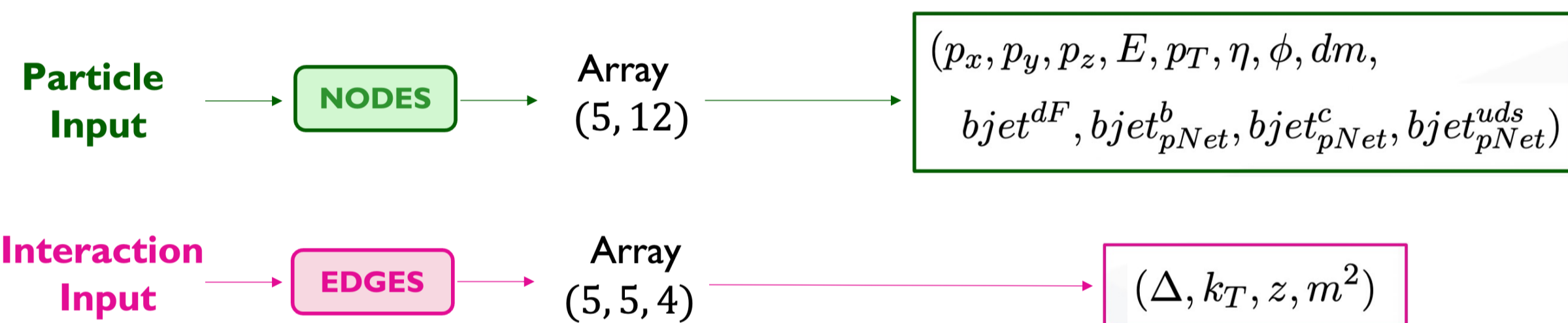
ParT [3]

Representation of the input

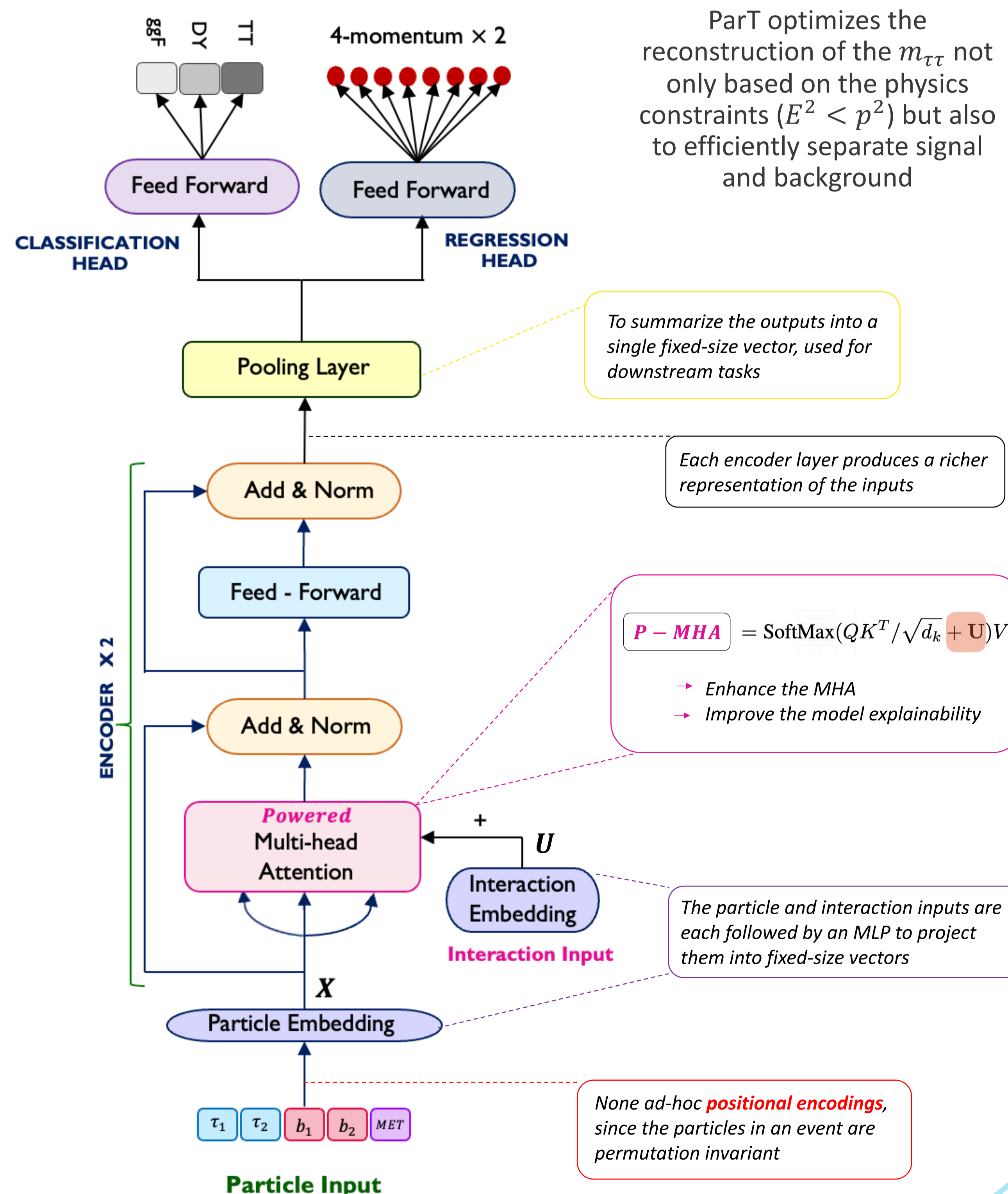
Each collider event has been modelled as a graph in which each node represents a visible decay product in the final state while the links correspond to the connection between them, described through pair-wise features



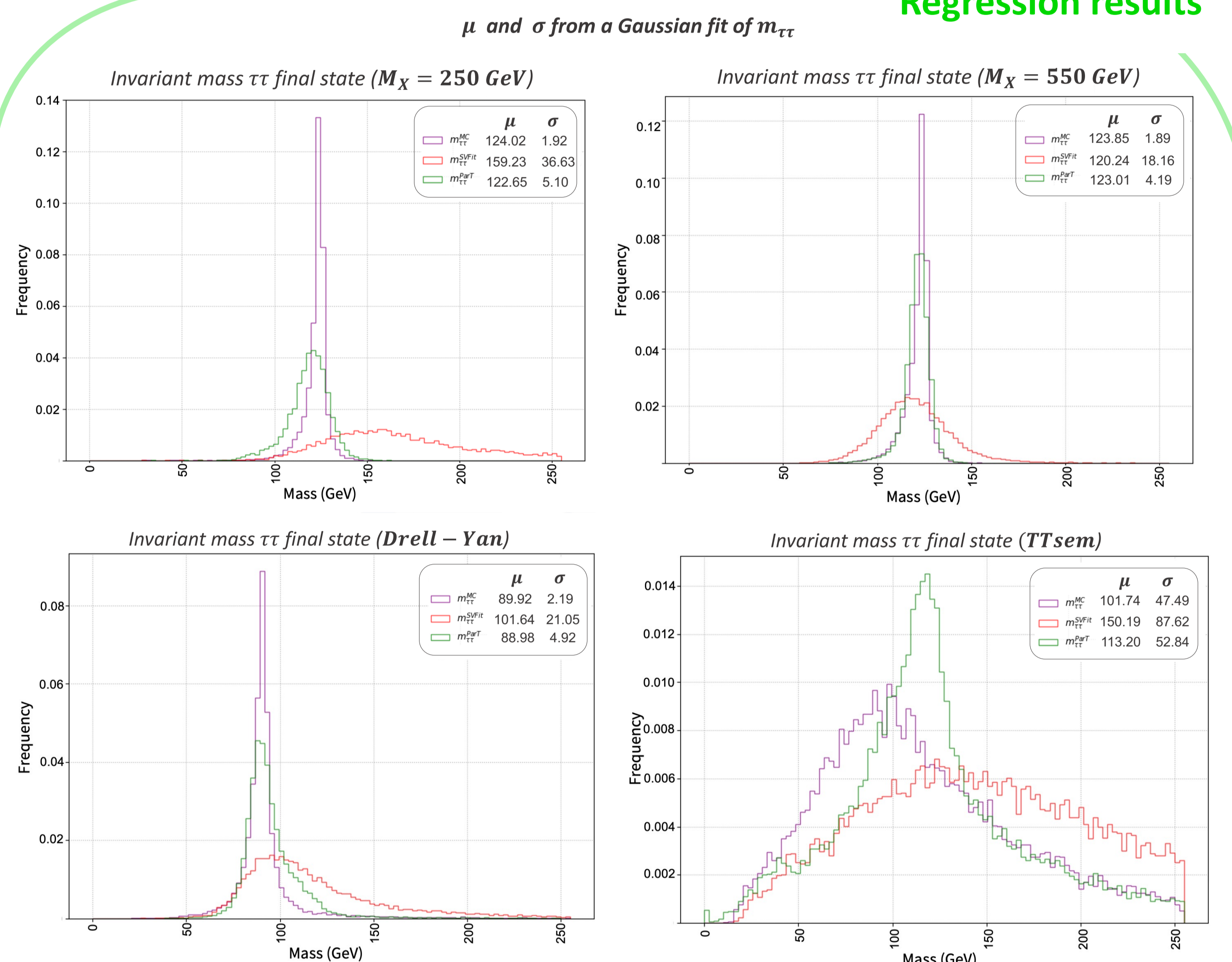
ParT is a Graph Neural Network (GNN)



The Model



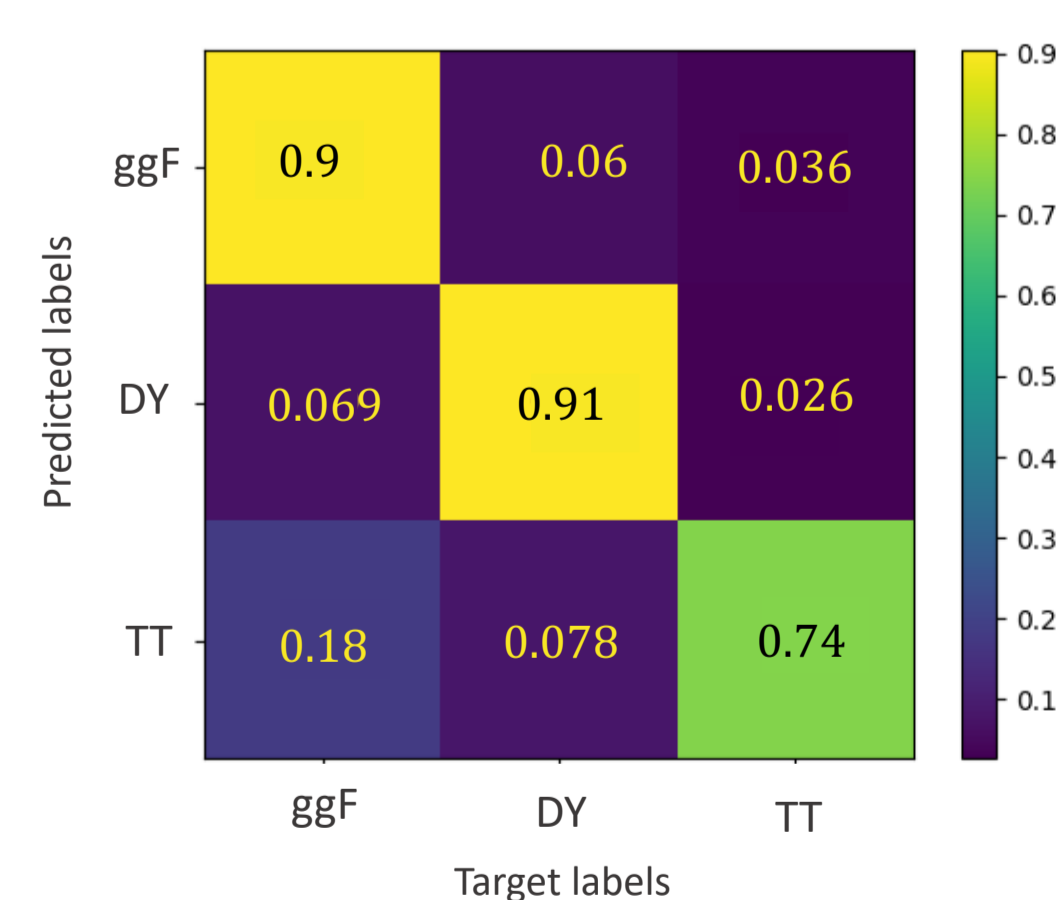
Regression results



With respect to SVFit performance:

- ✓ The mean value of the $m_{\tau\tau}^{\text{ParT}}$ distribution is much closer to the target one $\mu(m_{\tau\tau}^{\text{MC}})$ (computed combining the reconstructed τ leptons and gen-level neutrinos)
- ✓ The $m_{\tau\tau}^{\text{ParT}}$ distribution has a smaller standard deviation allowing for a more resolute $m_{\tau\tau}$ reconstruction
- ✓ ParT requires much less computation time: $O(10^{-3}s)$ vs. $O(1s)$ of SVFit

Confusion matrix



- The 90% and 91% of ggF and DY events, respectively, are correctly identified
- The $t\bar{t}$ sample is the one that is recognized with a lower precision, with a considerable % of events incorrectly classified as signal
- The accuracy in classifying signal events improves with the increasing of M_X

Conclusions

- The mass reconstruction of the di- τ system resulting from a Higgs boson decay plays an important role in searches of new massive particles decaying into two Higgs bosons
- New and computationally less expensive methods with respect to the current used SVFit algorithm for the $m_{\tau\tau}$ reconstruction are under study: the Transformer architecture has demonstrated to be a competitive approach
- To generalize the obtained results, it would be necessary to train ParT also for other production processes and the fully leptonic decay channels of the τ leptons

References

- CMS collaboration. "Search for nonresonant Higgs boson pair production in final state with two bottom quarks and two tau leptons in proton-proton collisions at $s = 13$ TeV." *Physics Letters, Section B: Nuclear, Elementary Particle and High-Energy Physics* 842 (2023): 137531.
- Bianchini, Lorenzo, et al. "Reconstruction of the Higgs mass in $H \rightarrow \tau\tau$ events by dynamical likelihood techniques." *Journal of Physics: Conference Series*. Vol. 513. No. 2. IOP Publishing, 2014.
- Qu, Huilin, Conggiao Li, and Sitian Qian. *arXiv: Particle Transformer for Jet Tagging*. No. arXiv: 2202.03772. 2022.