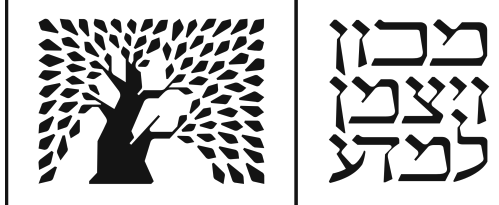
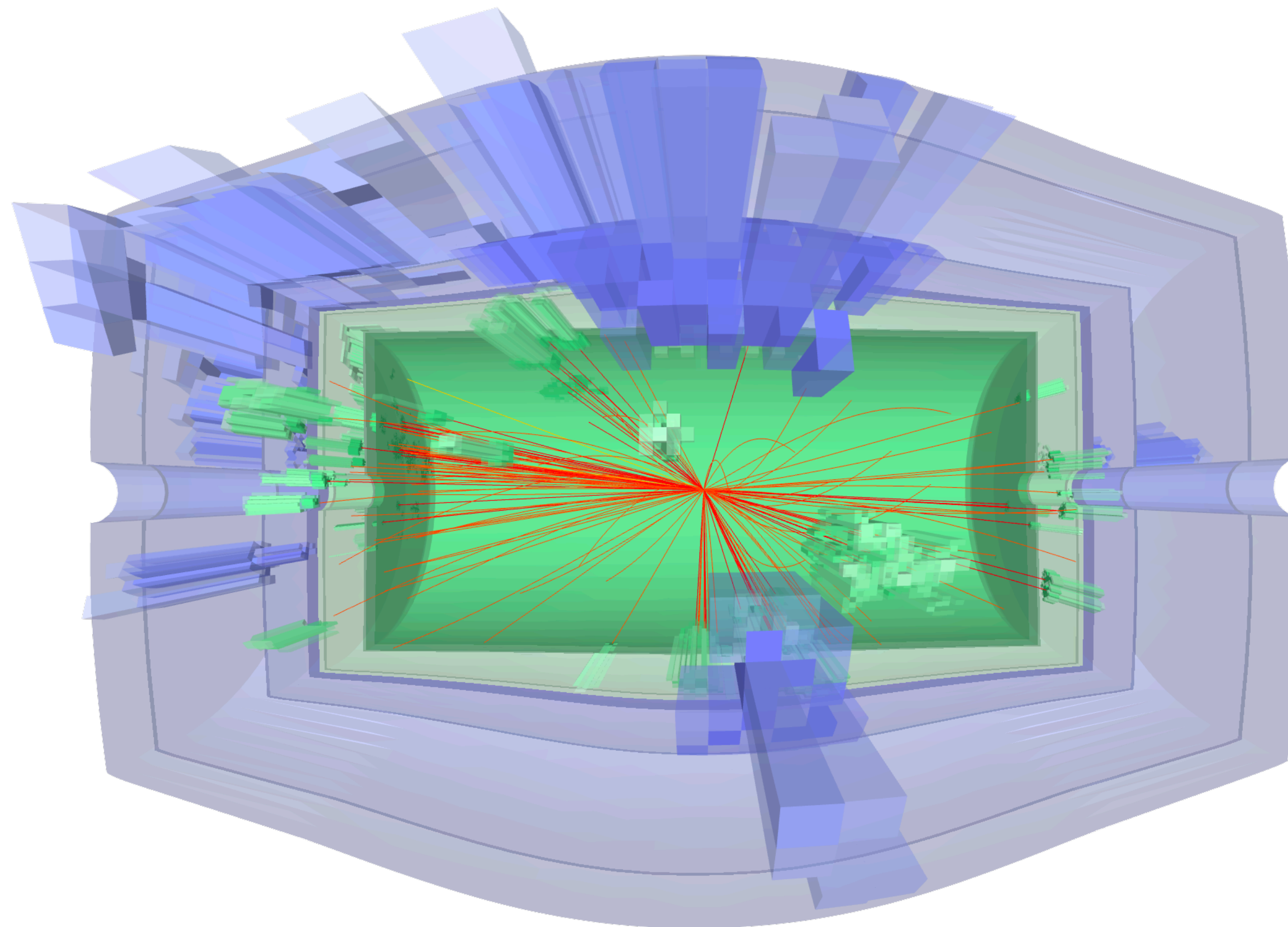
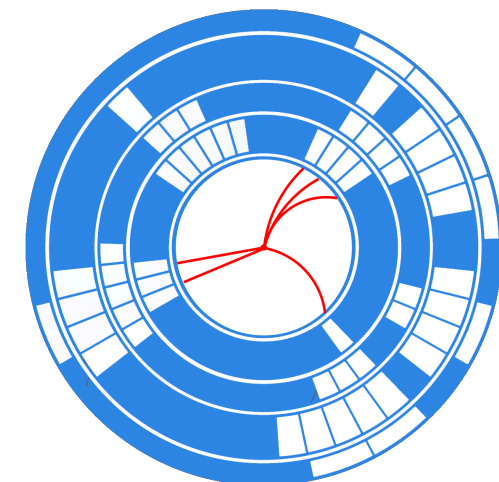


Novel Approaches for Fast Simulation in HEP using Diffusion and Graph-to-Graph Translation

E. Dreyer, E. Gross, N. Kakati, D. Kobylanski, N. Soybelman

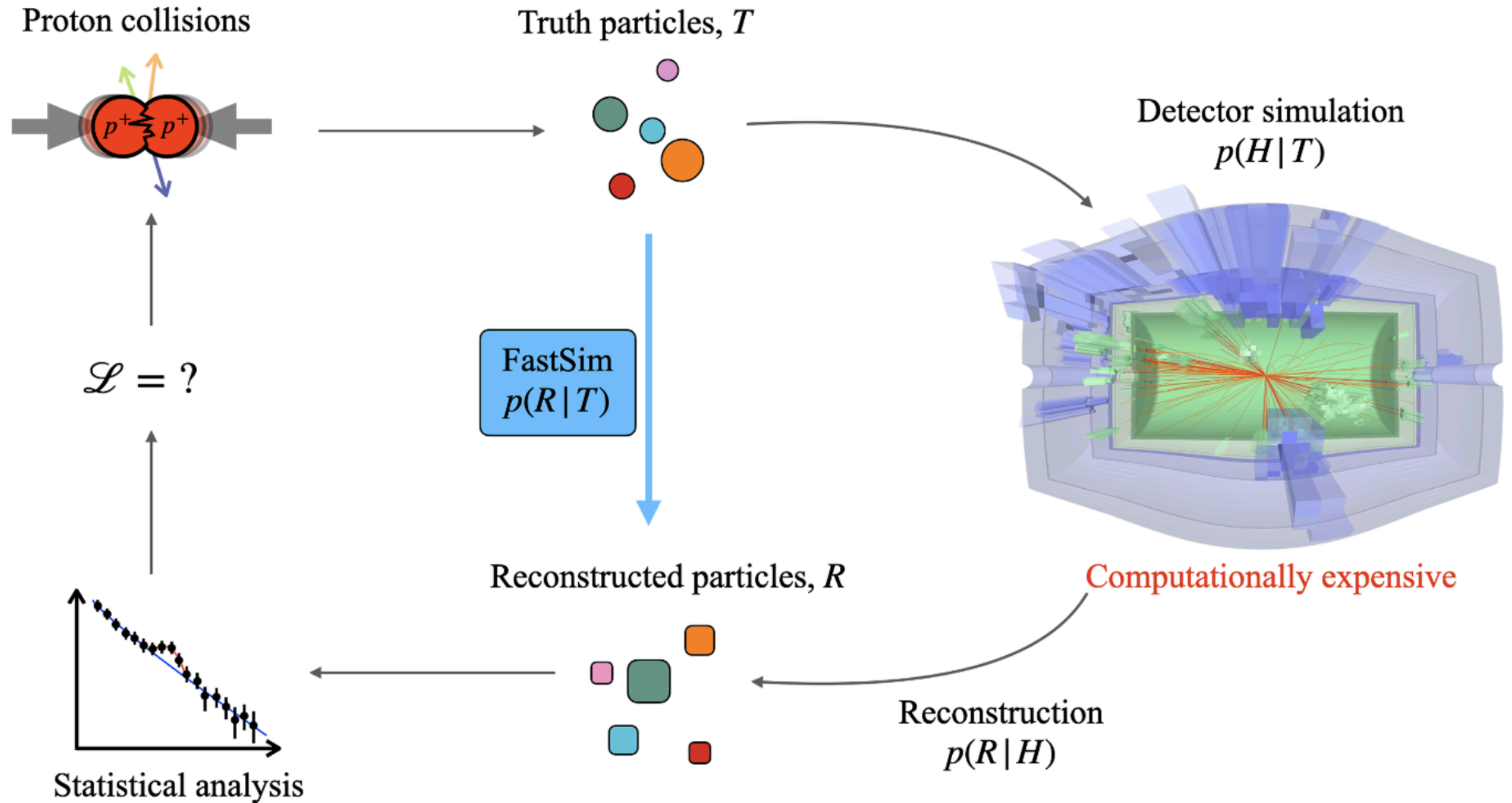


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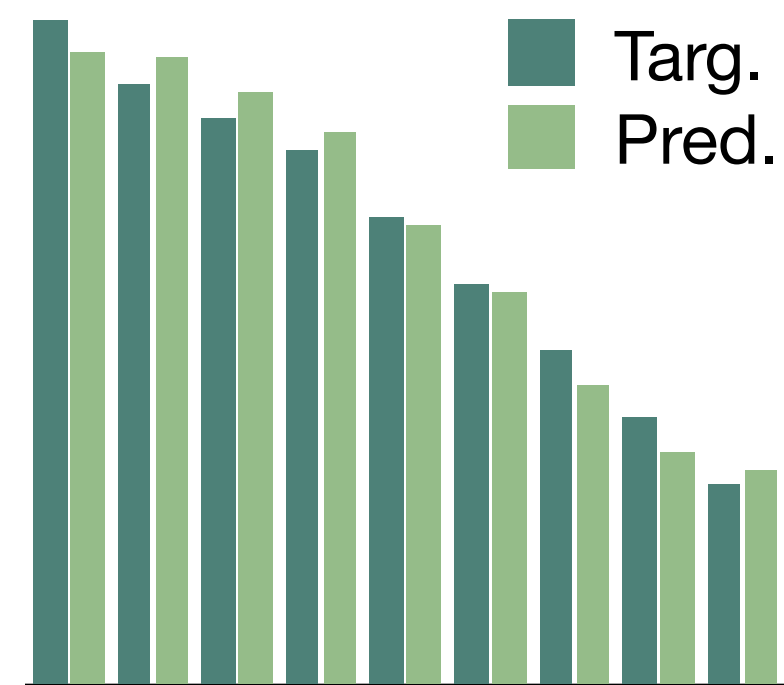
COCO

Motivation



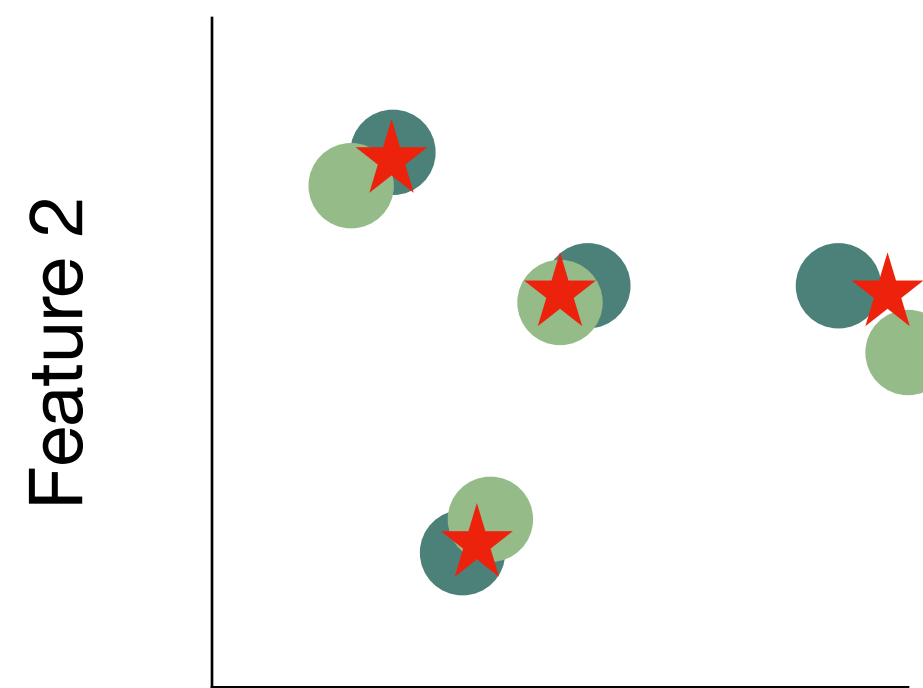
Goals

Marginal
distributions



Feature

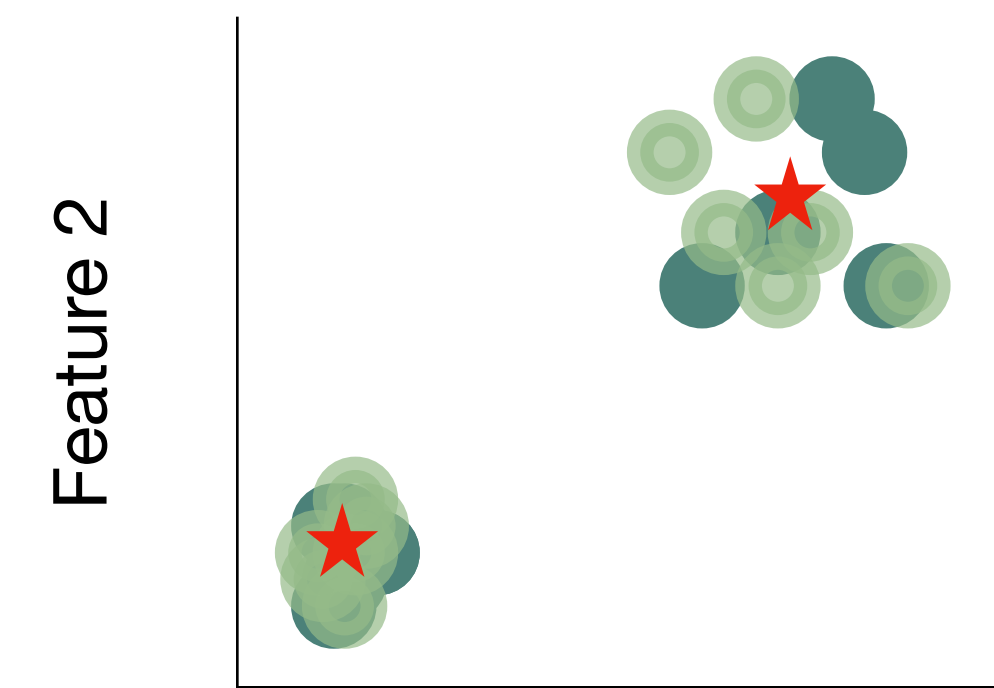
Reconstruct
constituents



Feature 1

3

Resolution

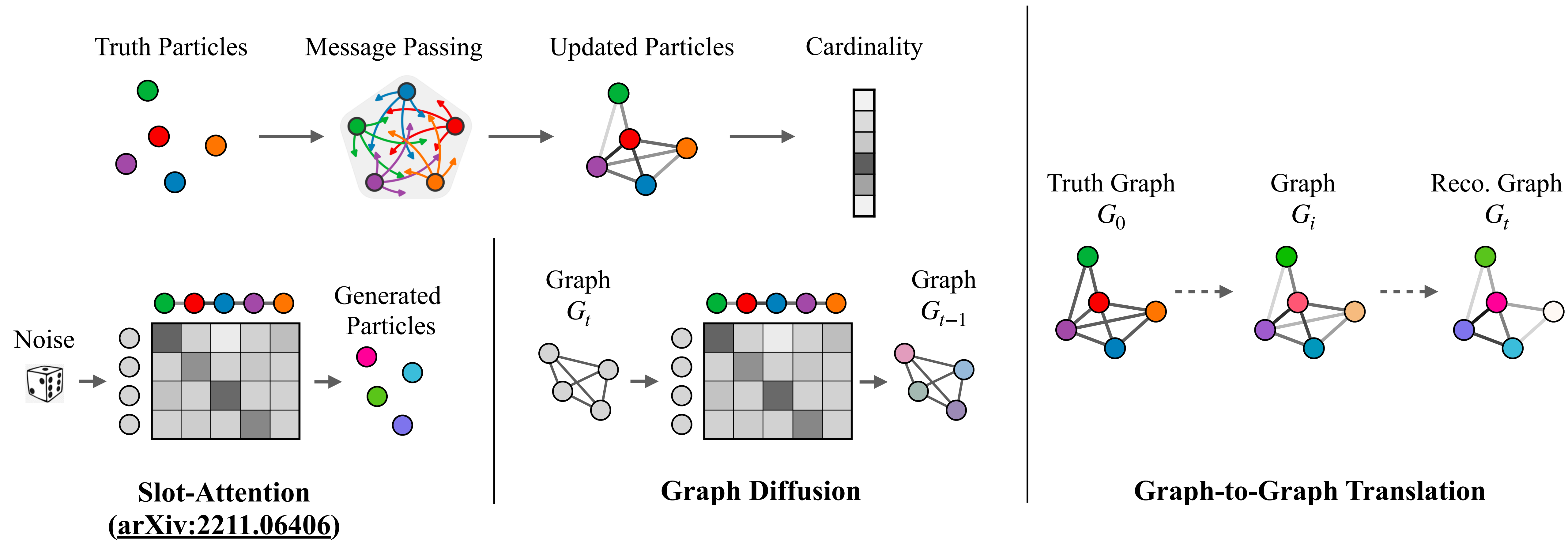


Feature 1

Dataset

- Single jet events generated with Pythia8
- Detector response simulated with COCOA ([arXiv:2303.02101](https://arxiv.org/abs/2303.02101))
- HGPflow reconstruction as target ([arXiv:2212.01328](https://arxiv.org/abs/2212.01328))
- Charged + Neutrals particles
- 100 replicas (multiple reconstructions of the same truth) per event to learn the smearing better

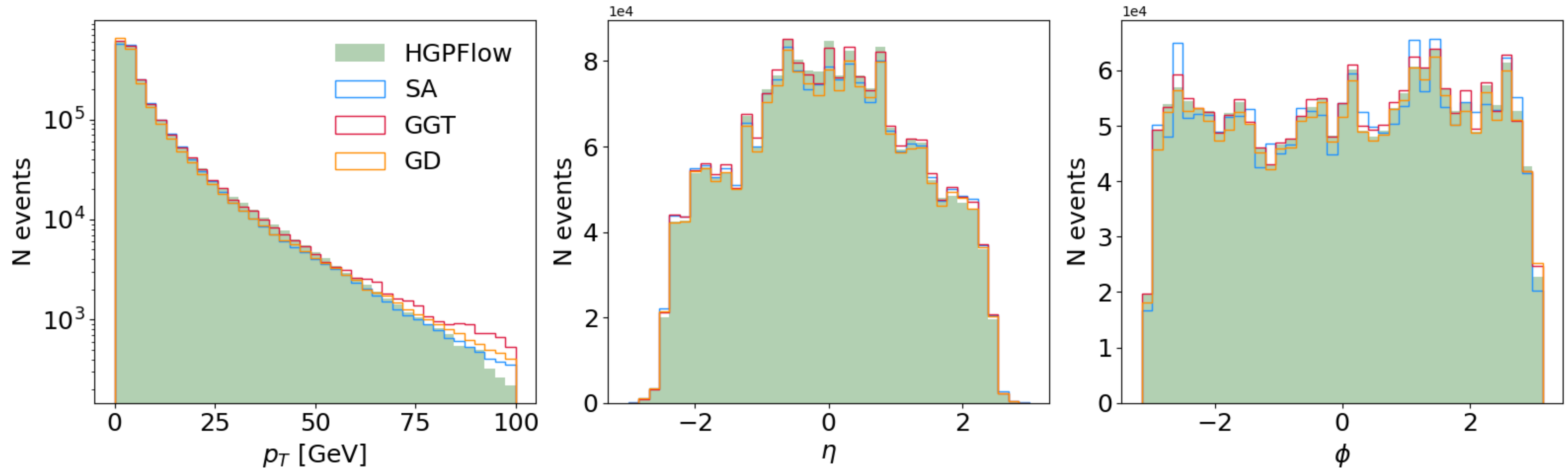
Architectures



For all models use MSE loss with Hungarian matching between target and prediction

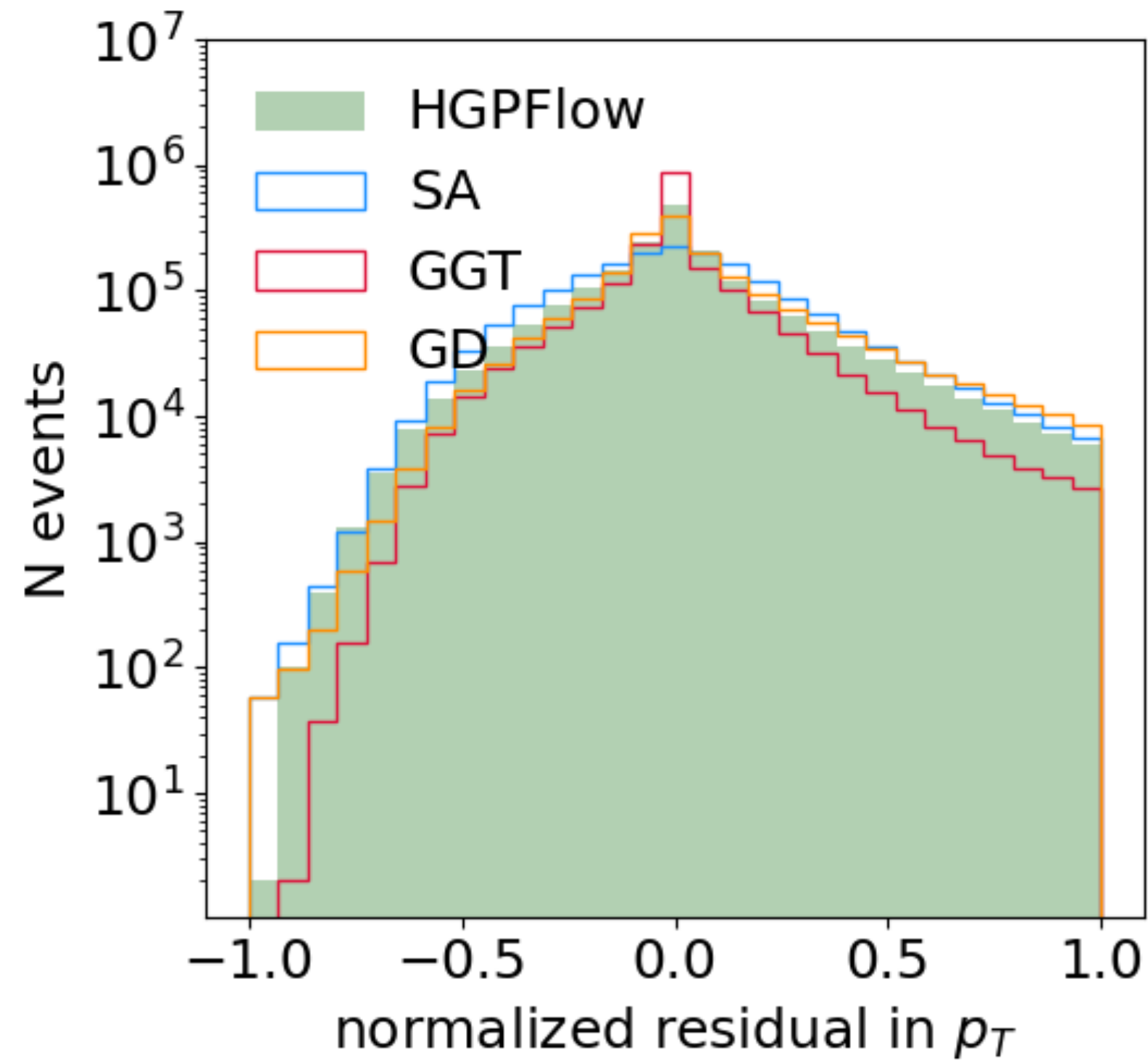
Results

Distributions: 3 features of particles (p_T, η, ϕ)

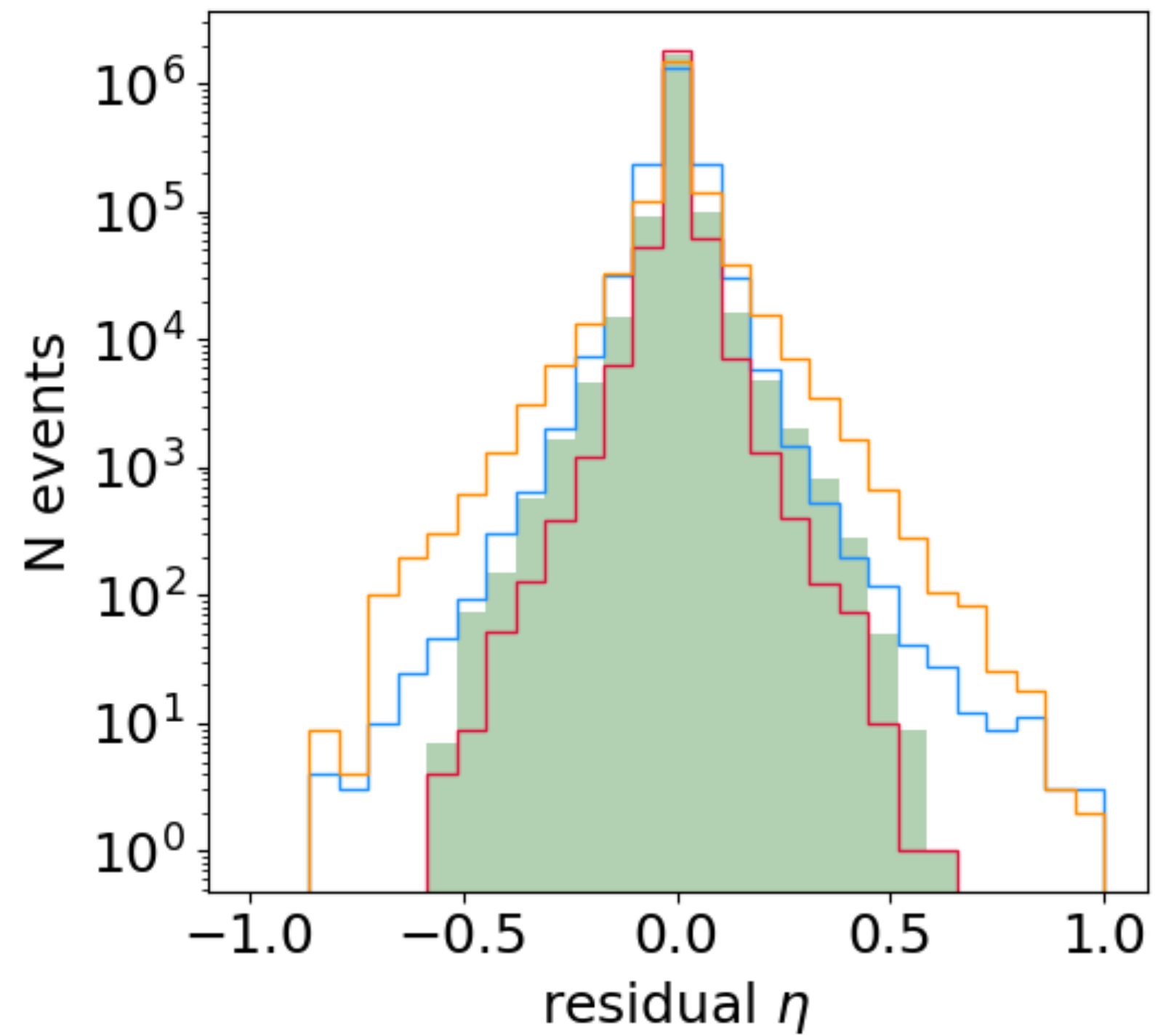


Results

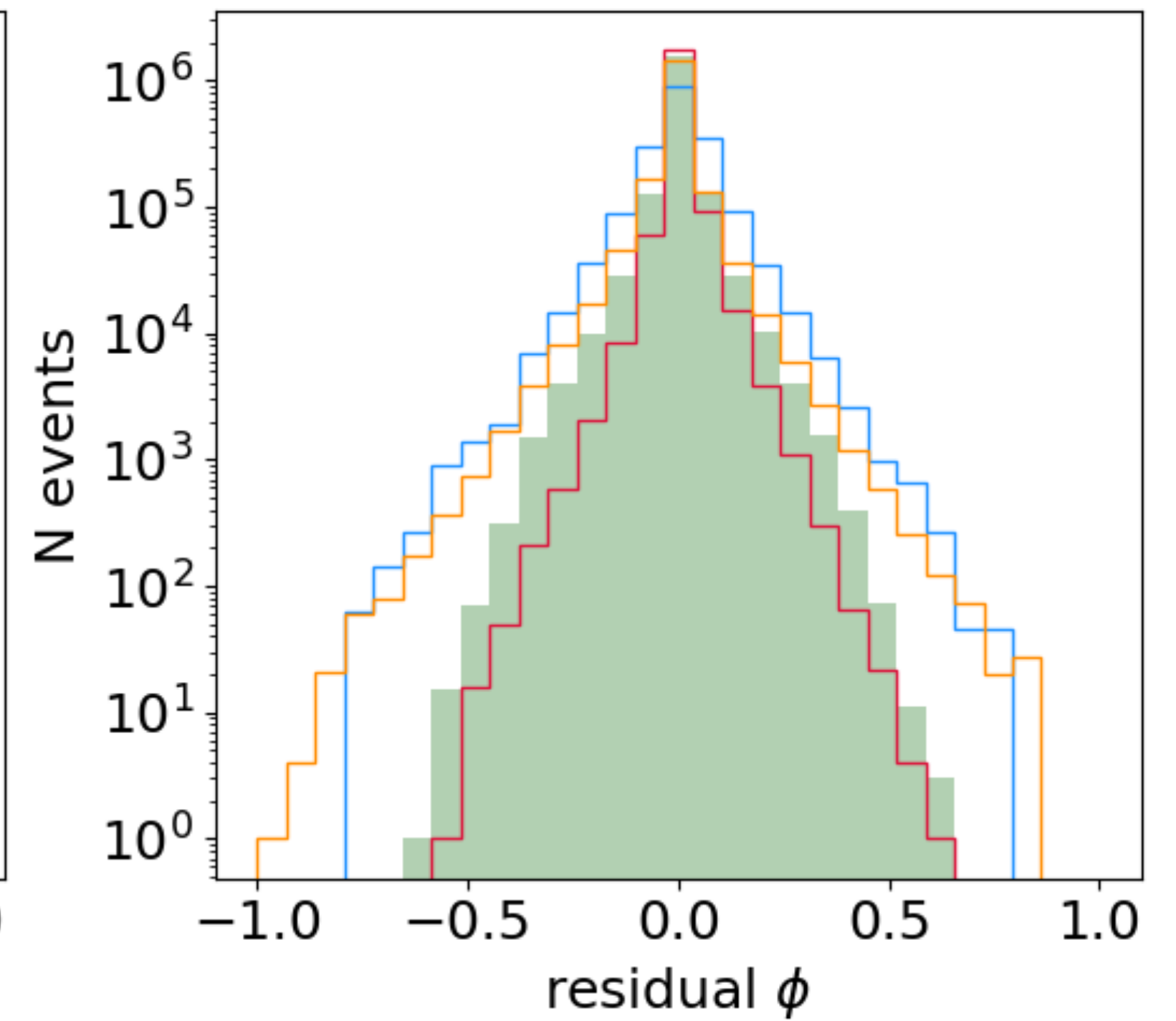
Residuals: 3 features of particles (p_T, η, ϕ)



$$\frac{p_{T,truth} - p_{T,reco}}{p_{T,reco}}$$

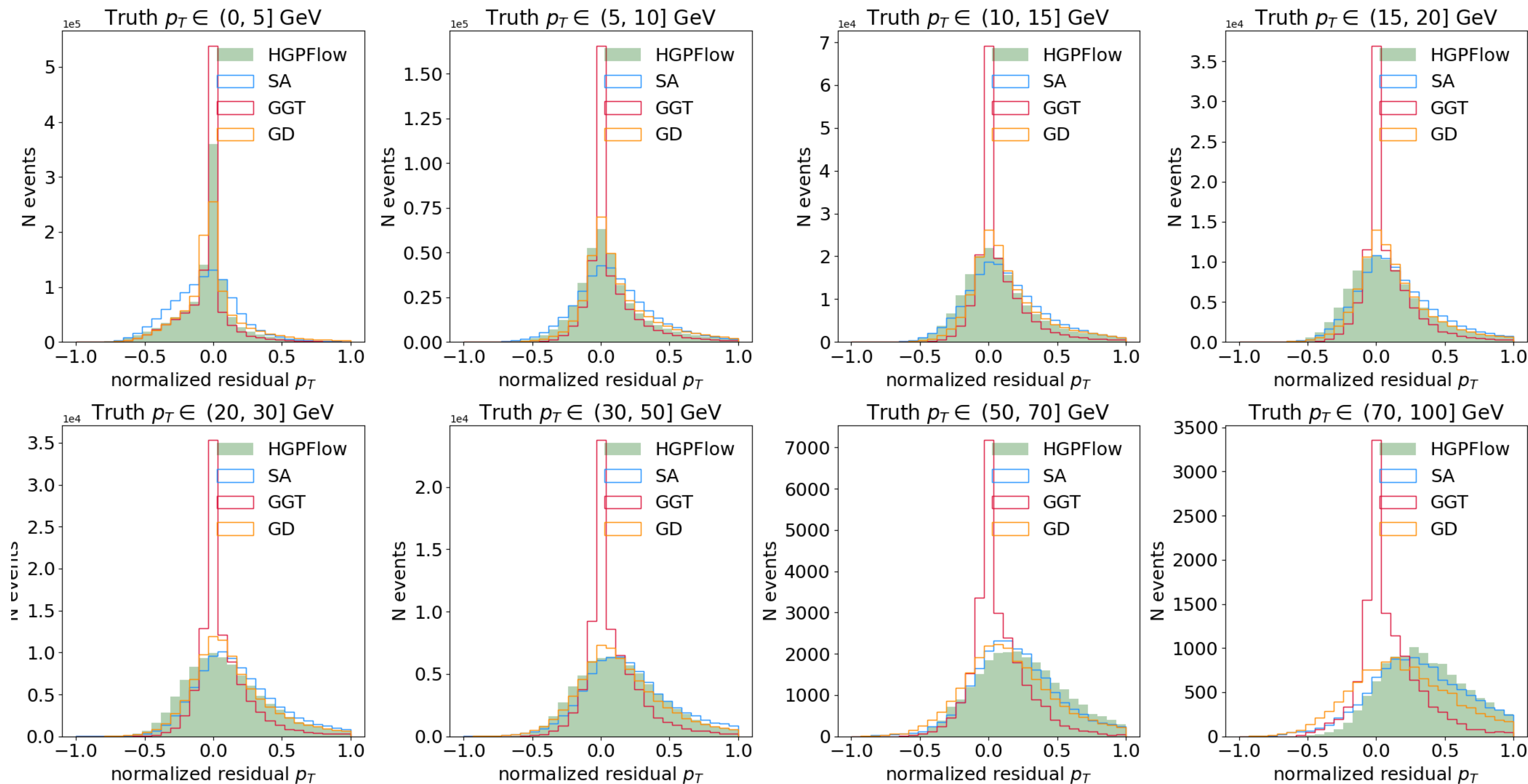


$$\eta_{T,truth} - \eta_{T,reco}$$

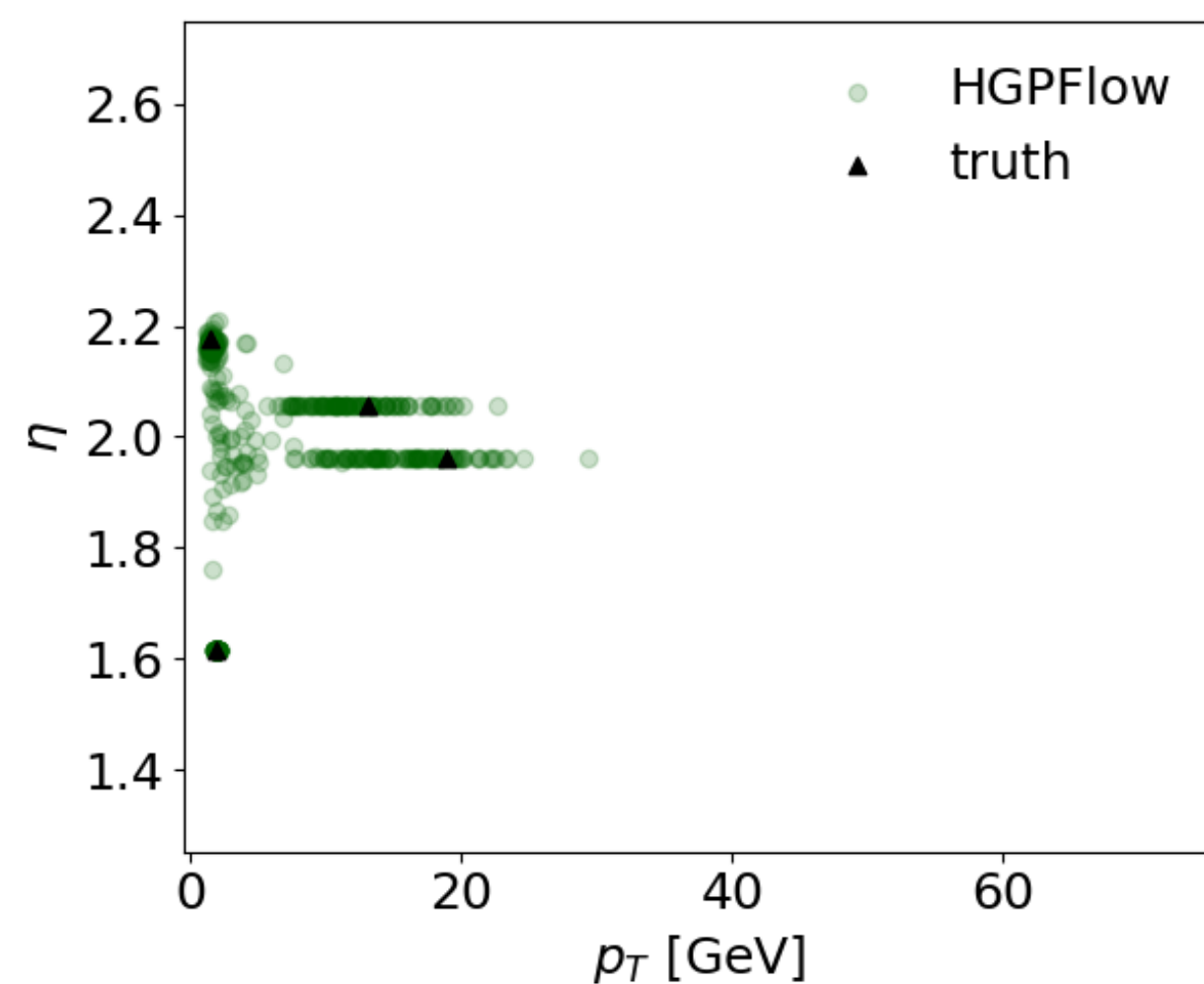


$$\phi_{T,truth} - \phi_{T,reco}$$

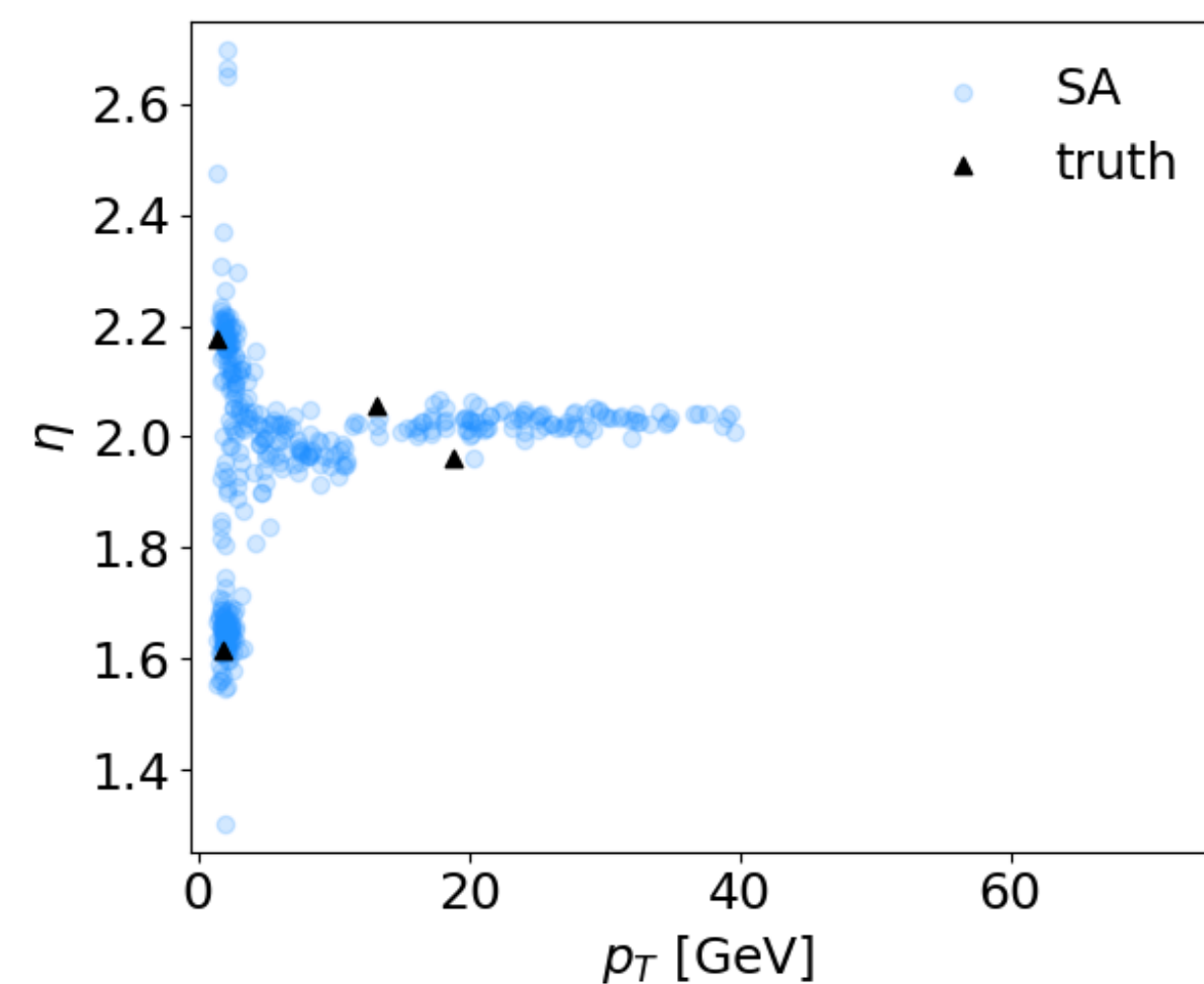
Residuals in p_T



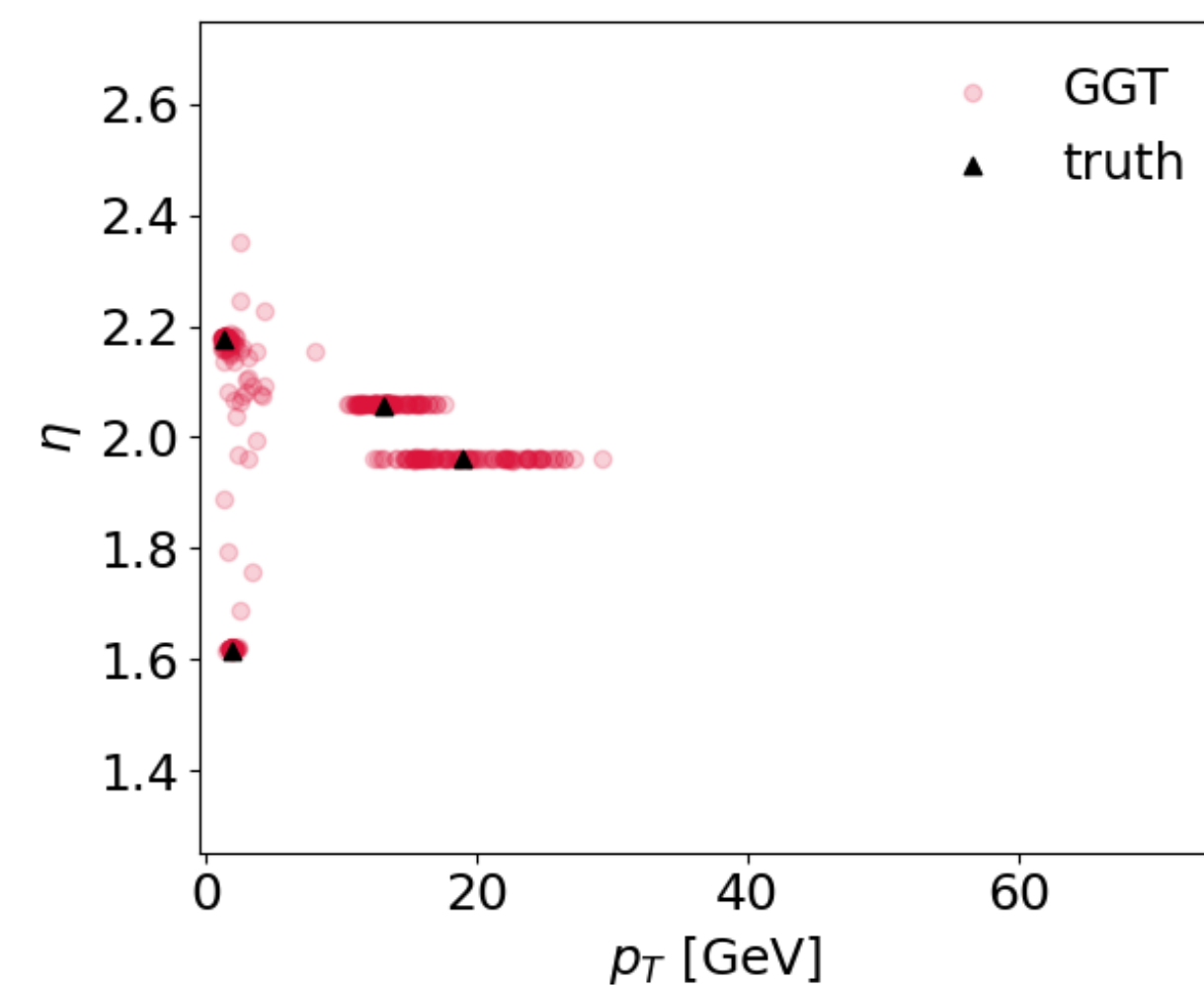
Example event displays



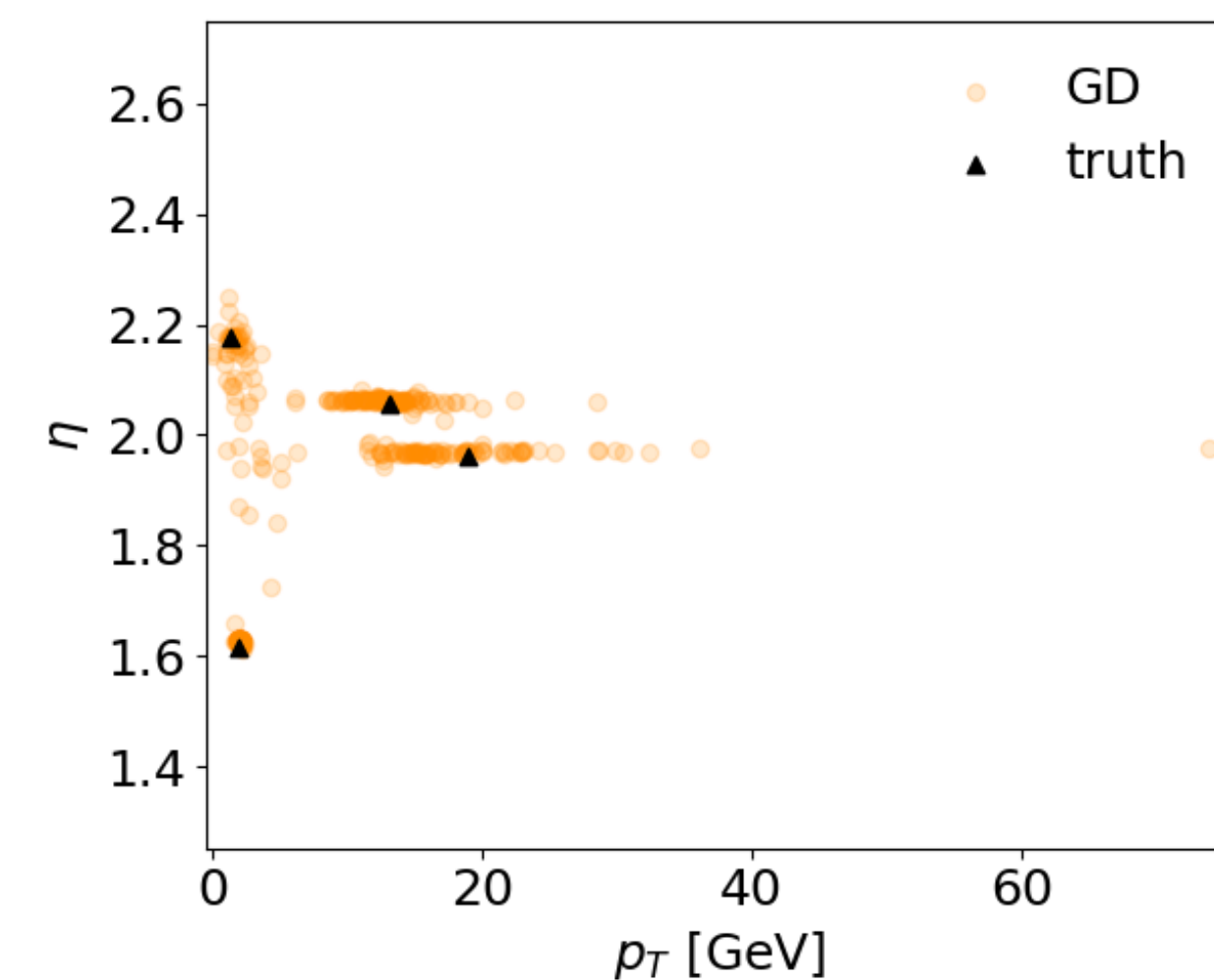
Target



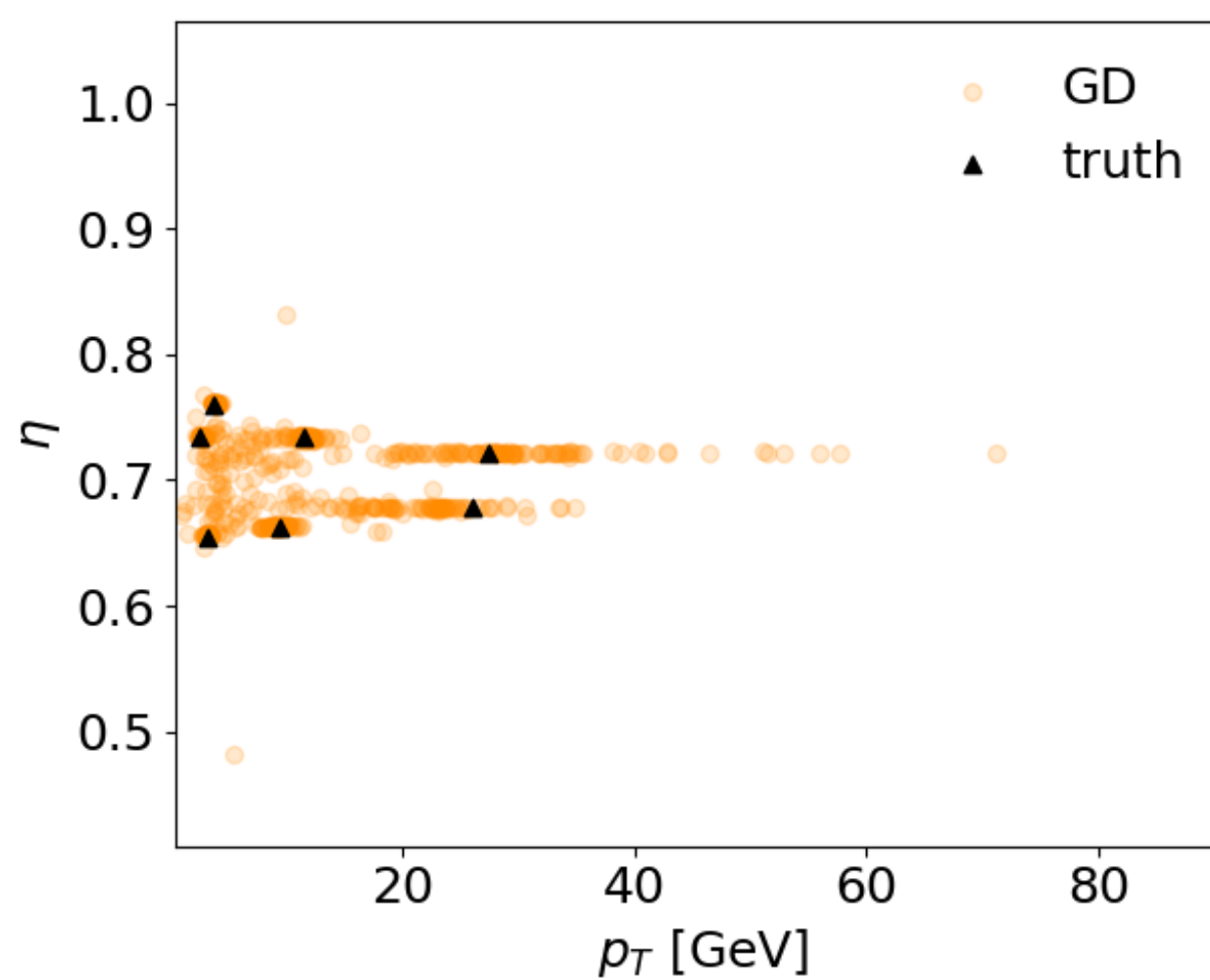
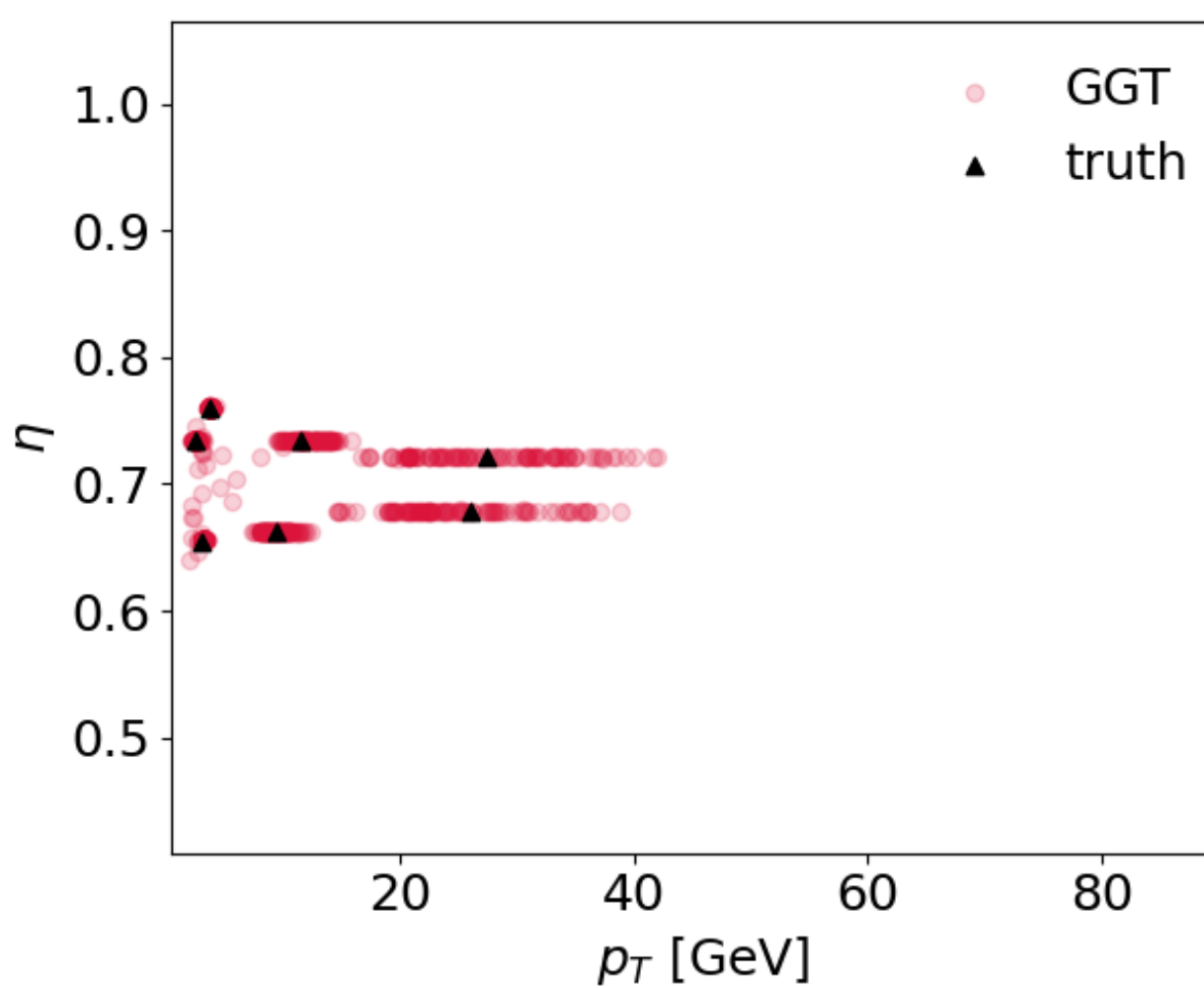
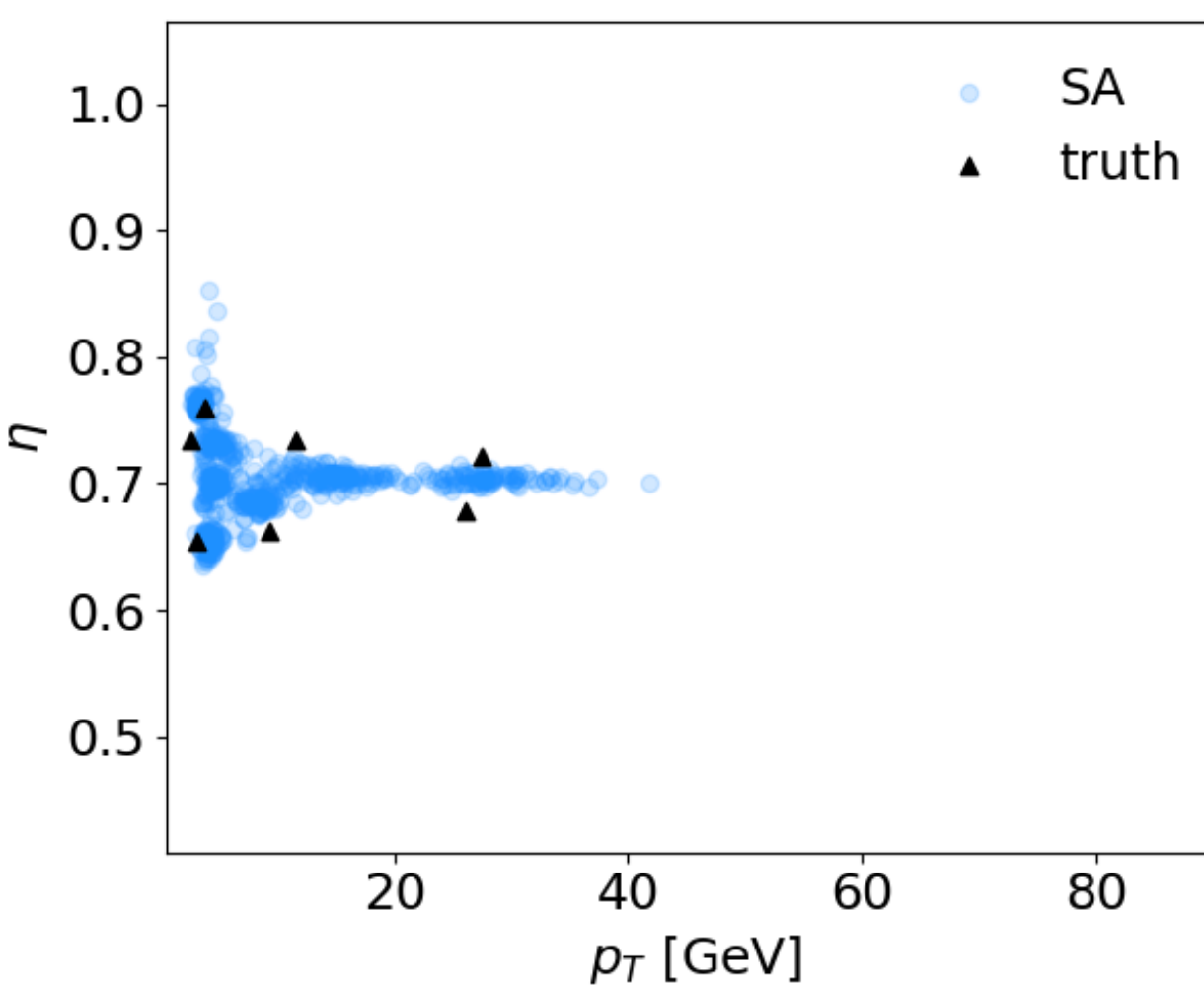
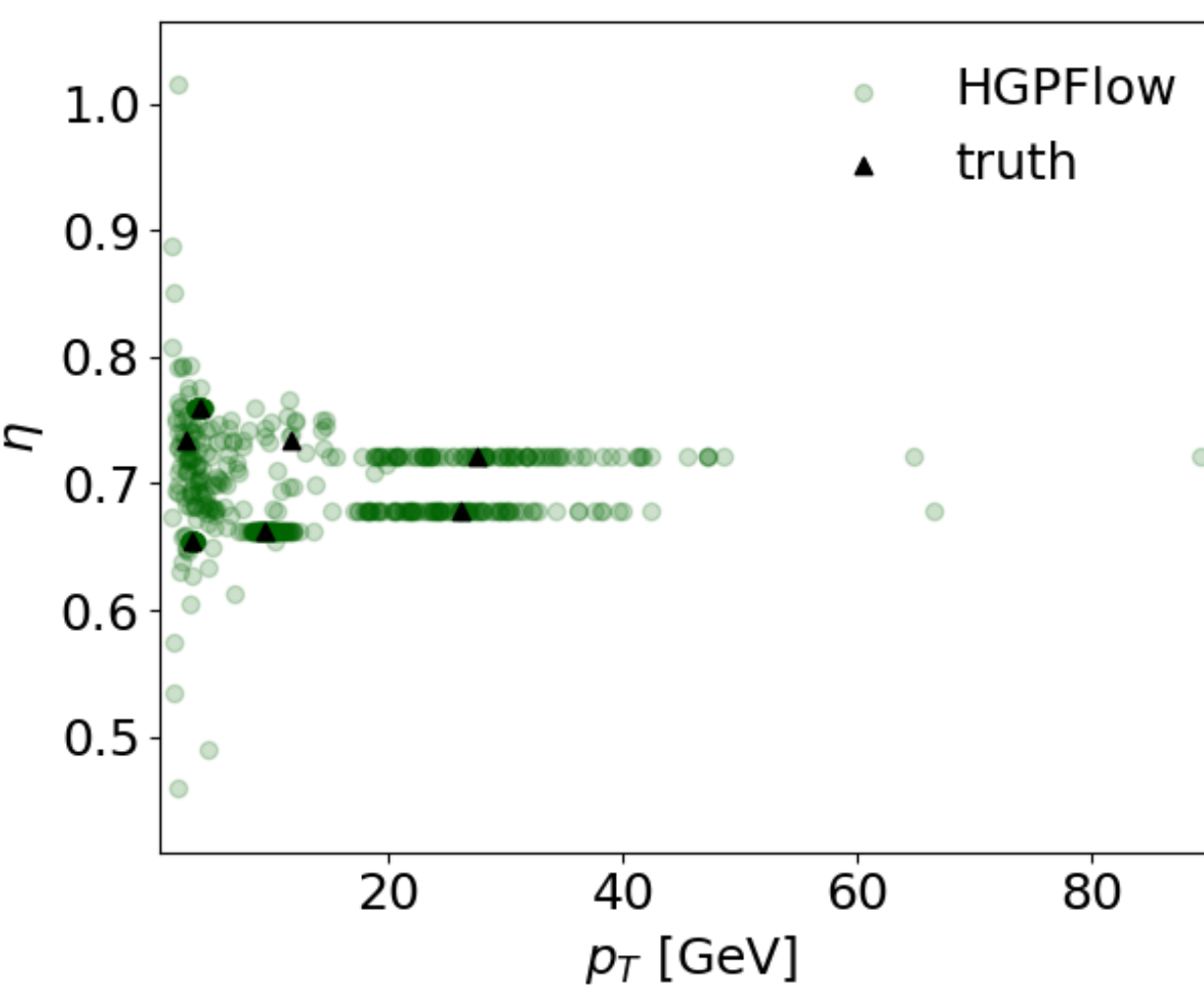
SlotAttention



Graph-to-Graph
Translation



Graph Diffusion



Conclusion

- Neutrals and charged particles can be well simulated by neural networks
- Both Graph-to-Graph Translation and Graph Diffusion outperforms baseline

Slot Attention model

- Graph-to-Graph Translation underestimates pT resolution, while Graph Diffusion captures it well