GNN LST Propagating the "cherry-on-top" GNN to LST efficiency plots April 4th, 2023

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- Exploring GNN for T5 classification
 - T5s have highest fake rate
 - Maybe GNN can improve over Balaji's x² "magic" cuts
- In the graph: lacksquare
 - T3s become "nodes"
 - T5 candidates become "edges"





Trial 1: GNN NTuple

- Training on T5s/T3s without duplicate removal
- pLS/pTX objects are excluded from the TCs
 - i.e. only T5s in TCs
 - Duplicate removal is applied here
- All cuts applied in T5 selection algo.
- Target question:

Can the GNN give us anything for free?

Trial 1: All cuts applied









Trial 1: GNN Performance ROC for LST master T5s, T5s in TC \rightarrow possibly large fake reduction!

- Solid curves show performance immediately after making χ^2 cuts
- Dashed curves show performance on only T5s in TC collection (i.e. after duplicate removal)
 - Q: If we run inference on only final T5s, **how** many fakes can we remove (while keeping X% of reals)?
 - A: We can **reduce fake T5s by > x2** while keeping 98% of reals
- **To-do:** propagate GNN score to LST NTuple and make DP-like efficiency plots







LST Master Efficiency Plots Showing p_T-binned eff. & FR for the **40 testing events**







LST GNN Efficiency Plots Showing p_T-binned eff & FR for the 40 testing events (98% sig eff)

Efficiency of Track Candidate

Sample:PU200 Version tag:190580D N_{evt}:40 $|\eta| < 4.5$, $|Vtx_z| < 30$ cm, $|Vtx_{xy}| < 2.5$ cm, Particle:All, Charge:All









LST Master Efficiency Plots Showing η-binned eff & FR for the **40 testing events**

Efficiency of Track Candidate



Efficiency





LST GNN Efficiency Plots Showing η-binned eff & FR for the 40 testing events (98% sig eff)





Summary

- The GNN is working:

 - Almost no efficiency loss
- All plots available here: <u>GNN, Master, GNN vs. Master</u>
- Next steps:
 - Add anchor hit φ's as T3 features
 - Edit efficiency code to allow us to look at the "Trial 2" efficiency



• "Trial 1" GNN from last week (i.e. "cherry-on-top") indeed reduces fake T5 TCs



Backup







GNN Inputs

or, layer
δr, layer
Sr, layer
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S

Scaled such that all features $\in [0, 1]$











Message passing — Latent^N grap

GNN Configuration

	Symbol	Name	Description
	Фe	Message function	 Neural network 3 hidden layers 200 nodes per layer
	Φv	Readout function	 Neural network 3 hidden layers 200 nodes per layer
	f _e	Edge classifier	 Neural network 3 hidden layers 200 nodes per layer
	ρ	Aggregator	Sum
bh	l Cla	$f_e \qquad \qquad$	





GNN Internal NN Configurations





Message function



Readout function





Trial 1: GNN Performance Showing inference ROC curves before/after duplicate removal (DR)







LST Master Efficiency Plots Showing η -binned eff & FR for the **40 testing events**

Efficiency of Track Candidate



Efficiency



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LST GNN Efficiency Plots Showing η-binned eff & FR for the 40 testing events (98% sig eff)





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LST GNN vs. Master Plots Showing pT-binned eff & FR for master (black) vs. GNN (red, 98% sig eff)



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LST Master vs. GNN Plots Showing pT-binned eff & FR for master (black) vs. GNN (red, 98% sig eff)









LST Master vs. GNN Plots Showing pT-binned eff & FR for master (black) vs. GNN (red, 98% sig eff)











LST Master vs. r-z x² Plots Showing pT-binned eff & FR for master (**black**) vs. GNN (red, 98% sig eff)



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