

# AI-assisted Four Top Quark Reconstruction

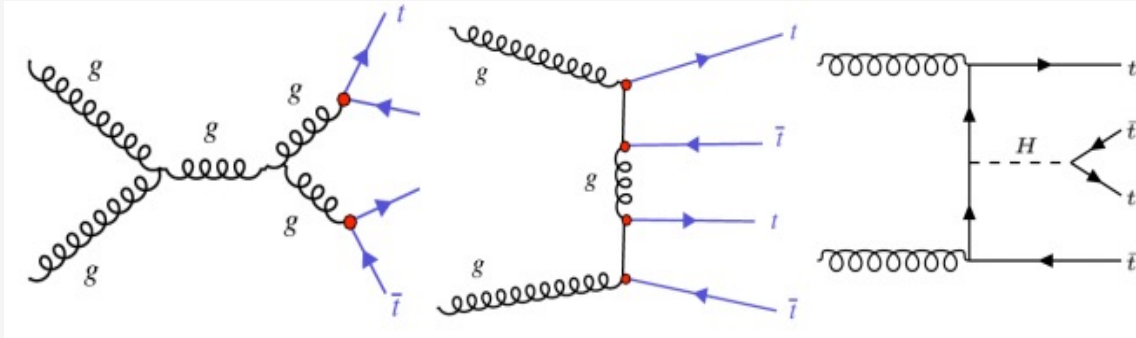
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# Rare Top Quark Production Process: Four-top-quark production

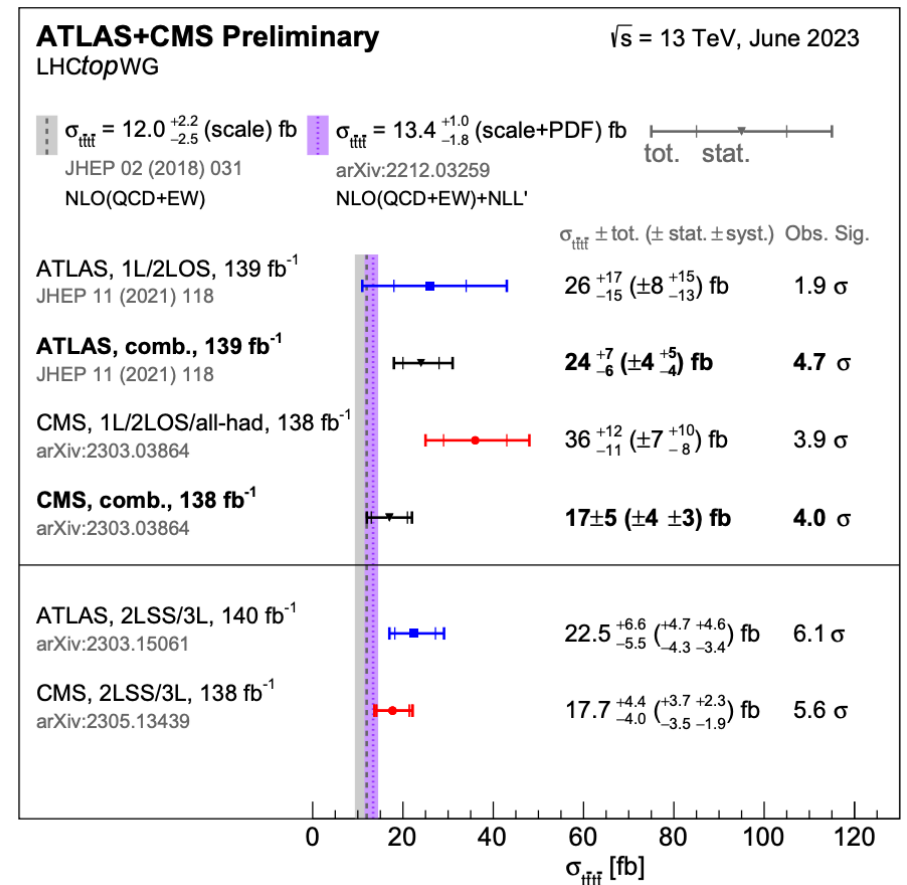
- $t\bar{t}t\bar{t}$  production is a rare top quark process predicted in the SM It is one of **the heaviest final states** accessible at LHC
  - NLO (QCD+EWK):  $\sigma(t\bar{t}t\bar{t}) = 12 \text{ fb} \pm 20\%$  [JHEP 02 (2018) 031]
  - NLO+NLL:  $\sigma(t\bar{t}t\bar{t}) = 13.4 \text{ fb} \pm 11\%$  [arXiv:2212.03259]



- $t\bar{t}t\bar{t}$  cross section is sensitive to anomalous top Yukawa coupling and Higgs CP properties
- Most sensitive channels in ATLAS and CMS are lepton channels: **SSML** and **1LOS**

[Eur. Phys. J. C 83 \(2023\) 496](#)

The observed significance is **6.1** sigma



# Top Quark Reconstruction

- Top quark reconstruction relies on the measurement of final states and jet-parton assignment is important in heavy particle reconstruction.
- Traditionally we use the  $\chi^2$  **method** to build each possible permutation of the event to find the best solution
  - Large number of permutations and long running time
  - Hard to reconstruct leptonic top
  - Depending on b-tagging

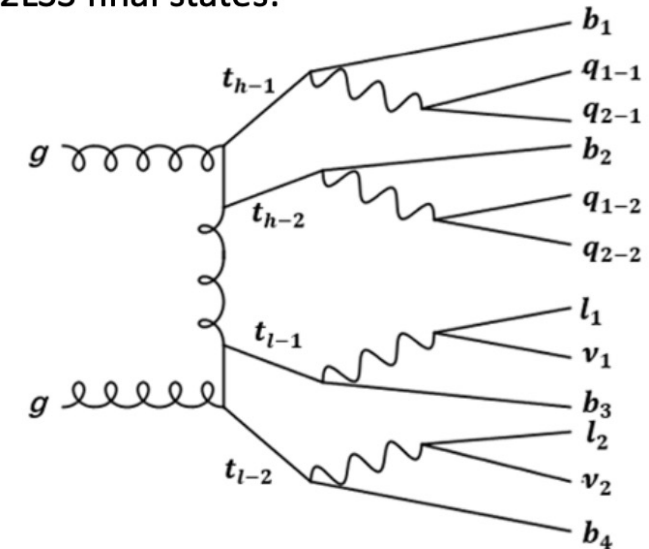
12 jets and 4 b jets: 2520 permutations  
14 jets and 5 b jets: 113400 permutations

Find the best permutation with smallest  $\chi^2$

$$\chi^2 = \frac{(m_{b_1 j_1 j_2} - m_{top})^2}{\sigma_{top}^2} + \frac{(m_{b_2 j_3 j_4} - m_{top})^2}{\sigma_{top}^2} + \frac{(m_{j_1 j_2} - m_W)^2}{\sigma_{m_W}^2} + \frac{(m_{j_3 j_4} - m_W)^2}{\sigma_{m_W}^2}$$

Event Type	N jet	Event fraction	Hadronic top chi2 efficiency
All Event	=6	7.9%	0.07
	=7	19.8%	0.130
	>=8	72.9%	0.1985
	inclusive	1	0.172

2LSS final states:



# Machine Learning Assignment Methods

Neural Network methods' advantages:

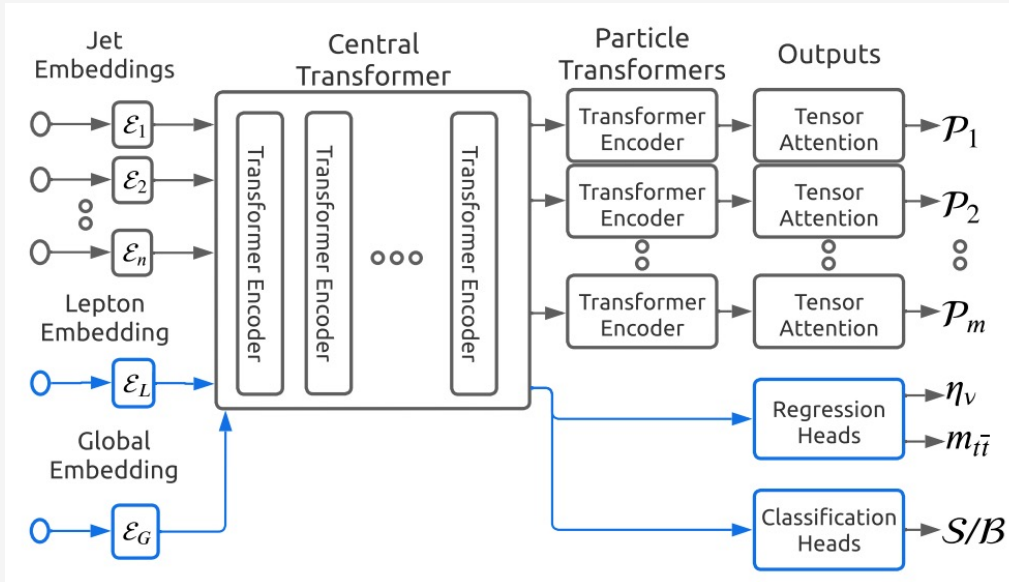
- Less time consumption in application: parallel computation by natural network
- High accuracy and can handle partial events

[Commun Phys 7, 139 \(2024\)](#)

[SciPost Phys. 12, 178 \(2022\)](#)

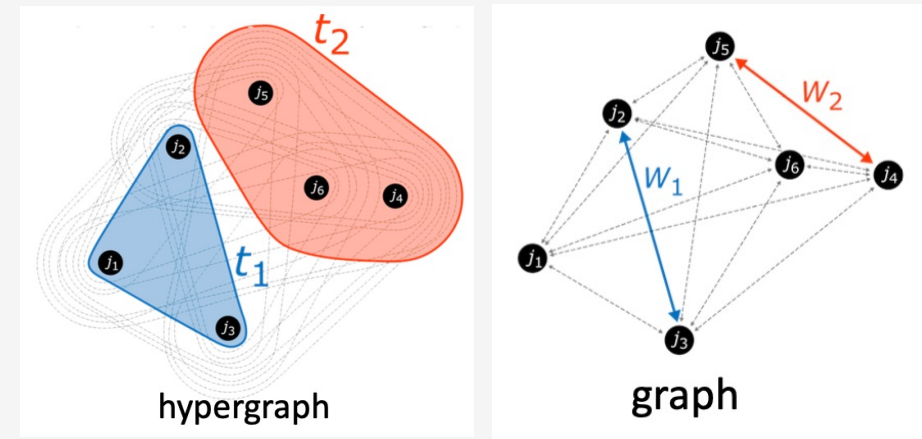
SPA-NET: use **transformer** to read multiple inputs from physics objects and output as sequence

$$Loss = \alpha_{reco} L_{reconstruction} + \alpha_{det} L_{detection} + \alpha_{reg} L_{regression}$$



[PhysRevD.111.032004](#)

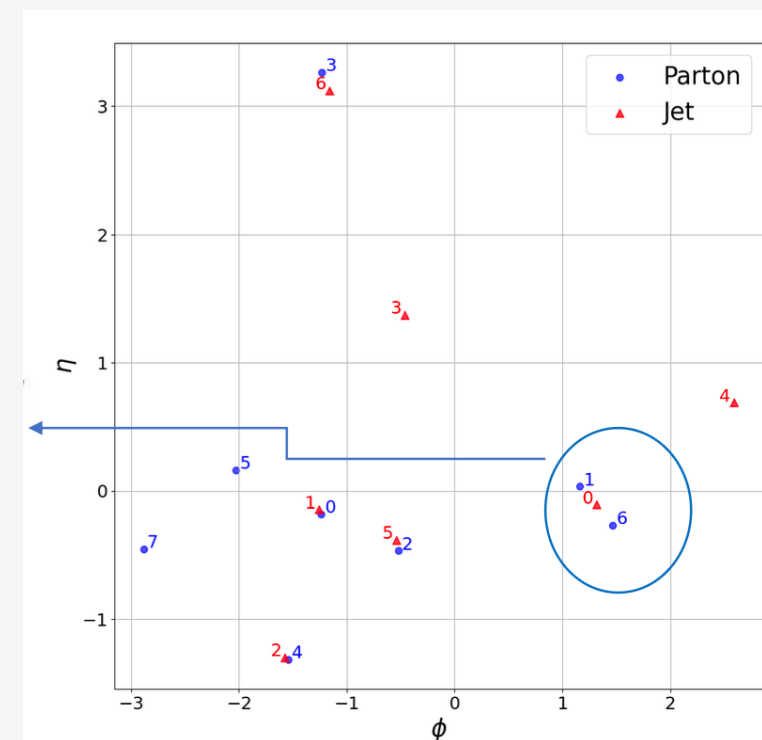
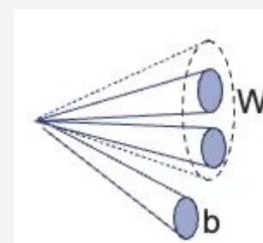
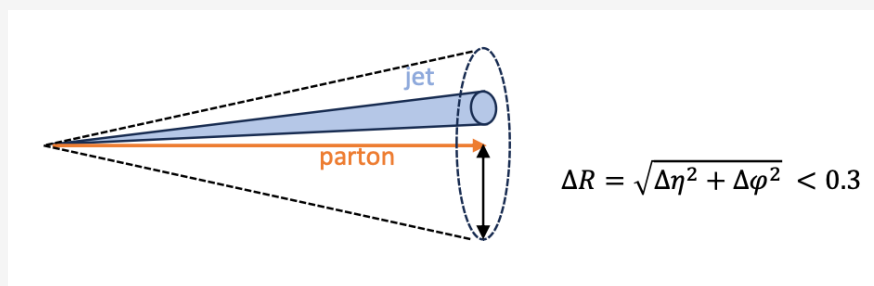
Hyper: use **hyper-graph** to group the top quark candidates in the graph dataset



# Truth Labelling for Supervised Learning

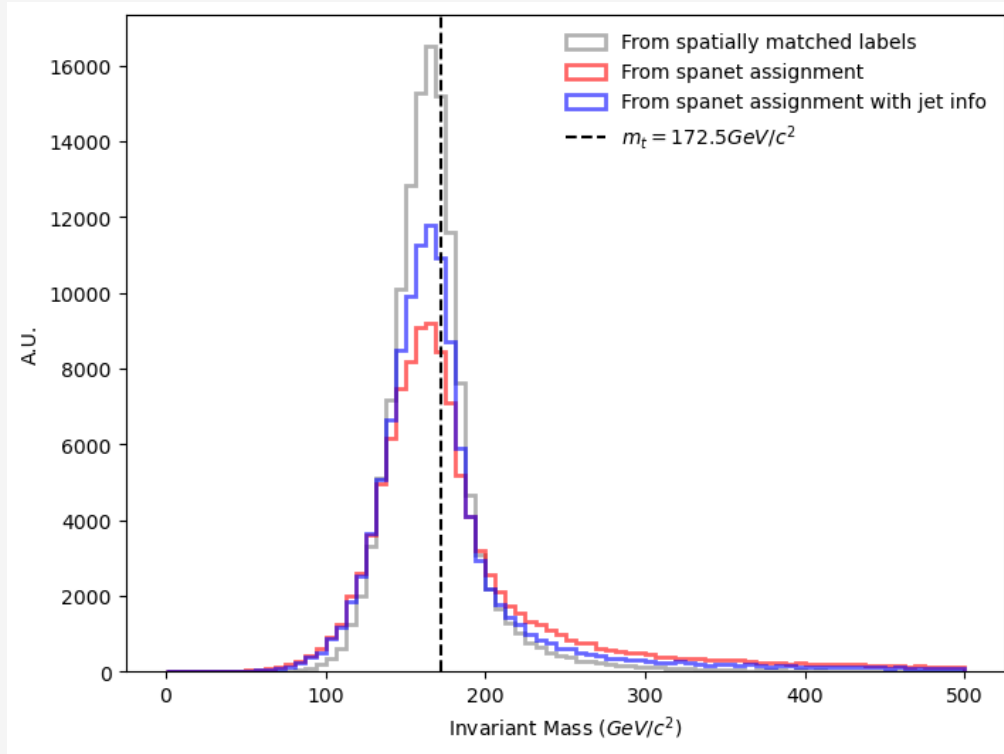
- Event generated by Madgraph5 and go through fast simulation using Delphes
- Surprised learning: need to label the jet in the final state
  - **Spatial matching** based on truth-level partons in (eta, phi) plane
  - Match the nearest jet to each parton: labelling efficiency is ~75%
- ~600k 4top sample in 2L channel for training and validation

Dataset	Number of events
Training	551k
Validation	29k
Evaluation	175k



# SPA-Net Performance in Four-Top-Quark Reconstruction

- Evaluating the performance using the top quarks which decay products are spatially matched to partons
- The machine learning method outperforms the traditional method in accuracy and speed:
  - Speed: SPA-Net improves about **3 times** faster in running time(both in CPU)
  - Accuracy: SPA-Net achieves **~38%** in hadronic top efficiency, **~70%** for leptonic top efficiency for four-top-quark events

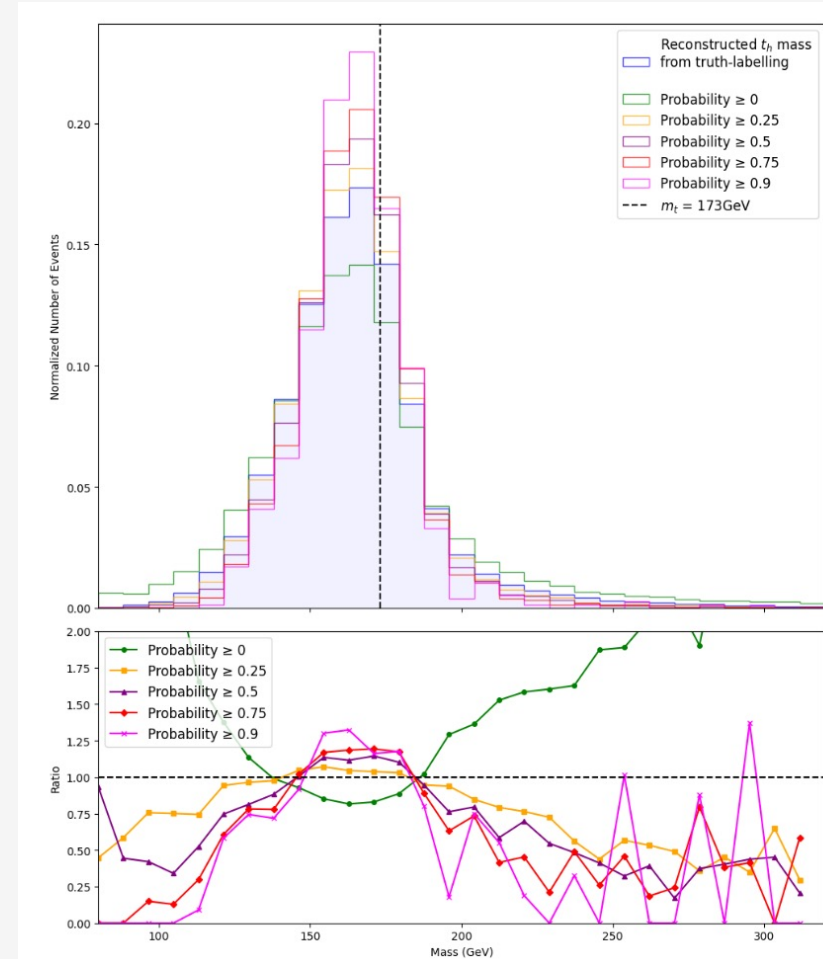
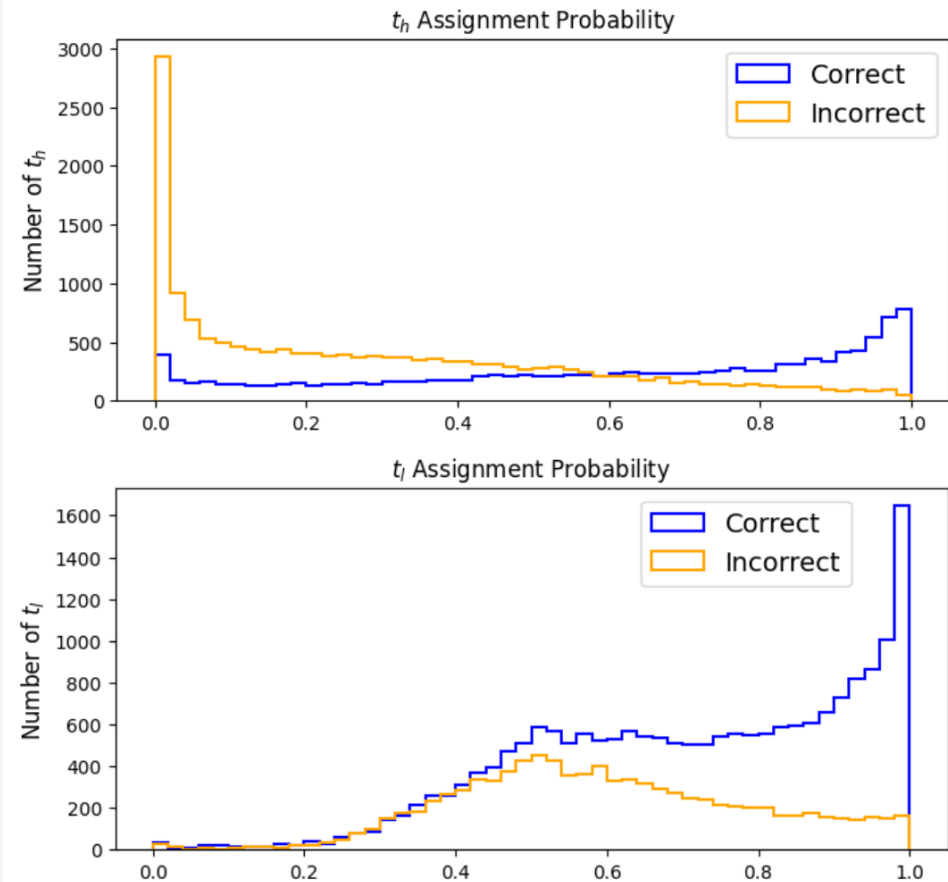


Even t Type	N jet	Event fraction	Hadronic top: $t_h$	Leptonic top: $t_l$	Hadronic top (add vars)	Leptonic top (add vars)
All Even t	=6	7.9%	0.326	0.752	0.332	0.766
	=7	19.8%	0.375	0.727	0.386	0.735
	>=8	72.9%	0.387	0.692	0.402	0.700
inclusive		1	0.383	0.703	0.397	0.711

- 200% improvement compared to  $\chi^2$  method**
- Adding  $\tau_1, \tau_2, \tau_3, \tau_4, \tau_5$  and ratio of charge particles over neutral particles (shown in brown column)
  - Small improvement (percent level)
- $t_l$  efficiency drops to 0.632 if require both b-quark and lepton

# SPA-Net Assignment Probability

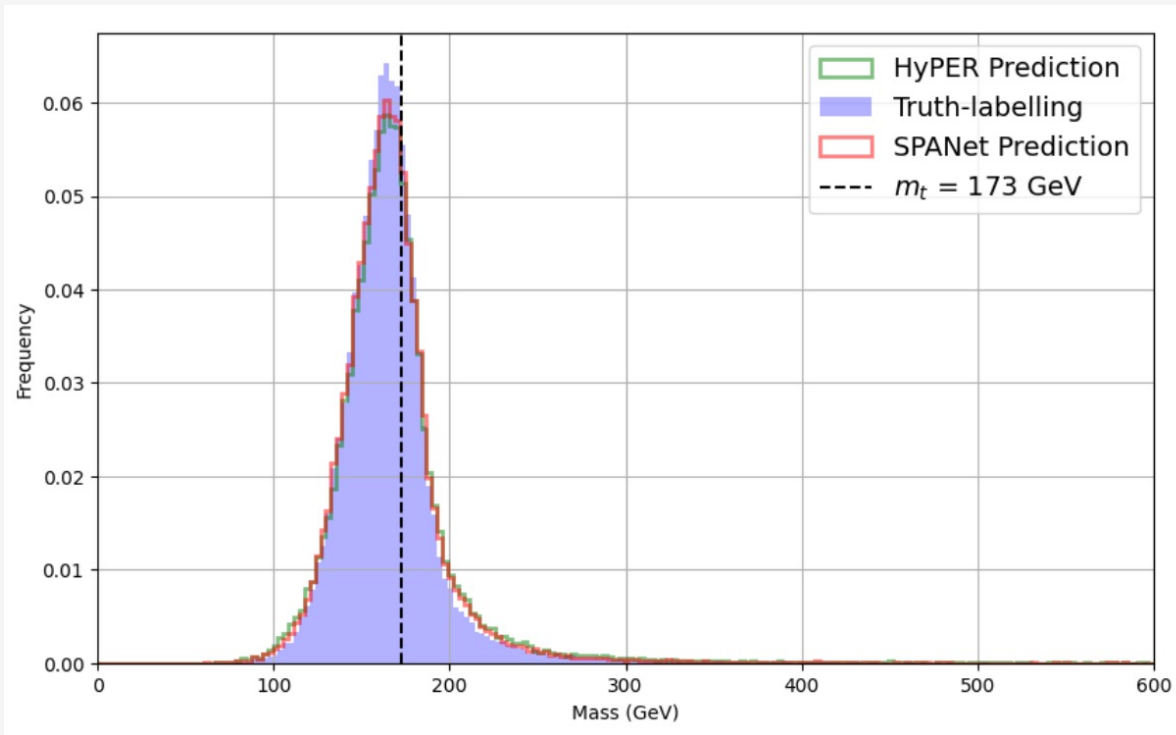
- The assignment probability can help to evaluate the output of the SPA-Net
  - Each top quark from SPA-Net prediction is classified as correct or incorrect by comparing it to truth label
  - Threshold on the assignment probability can help to improve the proportion of correct assignments





# Result of HyPER

- Hypergraph can derived similar results in hadronic top quark reconstruction
- Parameter-efficient compared to SPA-Net: 500k vs.16M
- Performance reaches **~47%** in hadronic top reconstruction



Event Type	N jet	SPANet (2 $t_h$ +2 $t_l$ )	SPANet (2 $t_h$ )	HyPER (2 $t_h$ )
All Event	=6	0.326	0.540	0.559
	=7	0.375	0.528	0.525
	>=8	0.387	0.461	0.453
	inclusive	0.383	0.471	0.467

- HyperGraph has no neutrino regression and can only reconstruct  $t_h$
- SPA-Net and HyperGraph have similar performances when reconstructing  $t_h$  only
- Adding neutrino regression should help for HyperGraph



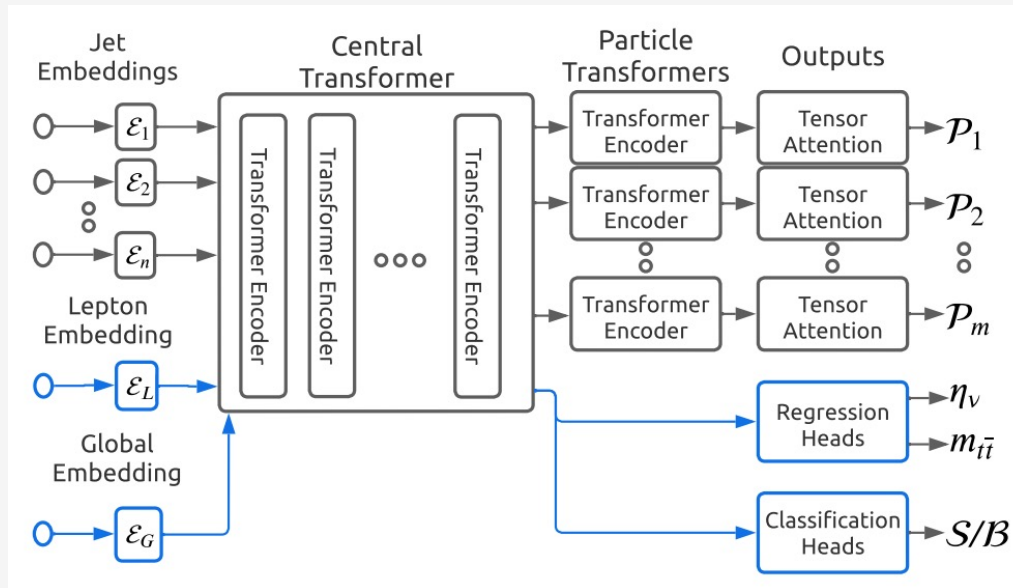
# Multi-task Learning in Multi-lepton Channel

- In two lepton channel, the large missing ET coming from the neutrino arise the difficulty for reconstruction
- Machine learning method can regress or predict the momentum of the neutrino

[Commun Phys 7, 139 \(2024\)](#)

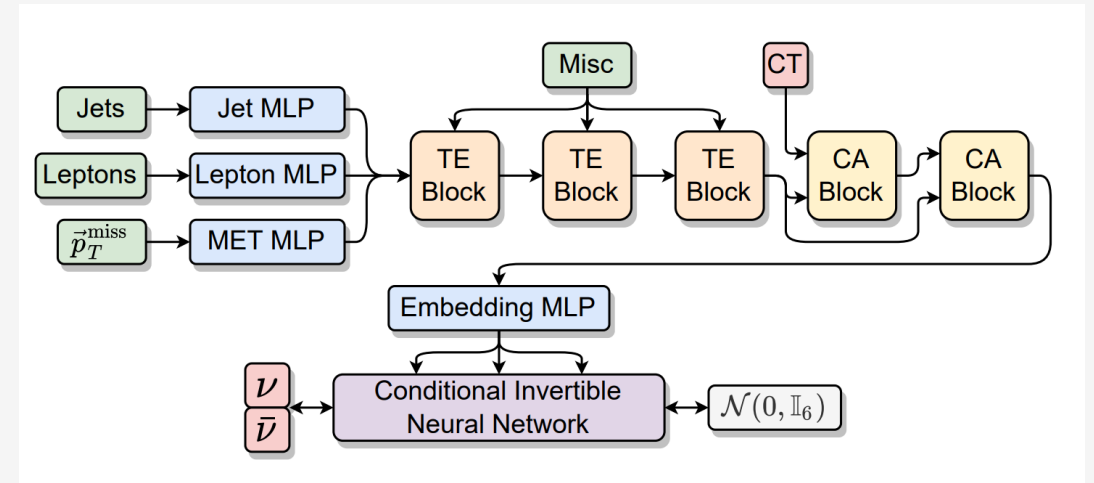
[SciPost Phys. 12, 178 \(2022\)](#)

- SPA-Net: use **regression** heads to output the value of the four momentum of neutrino



[PhysRevD.109.012005](#)

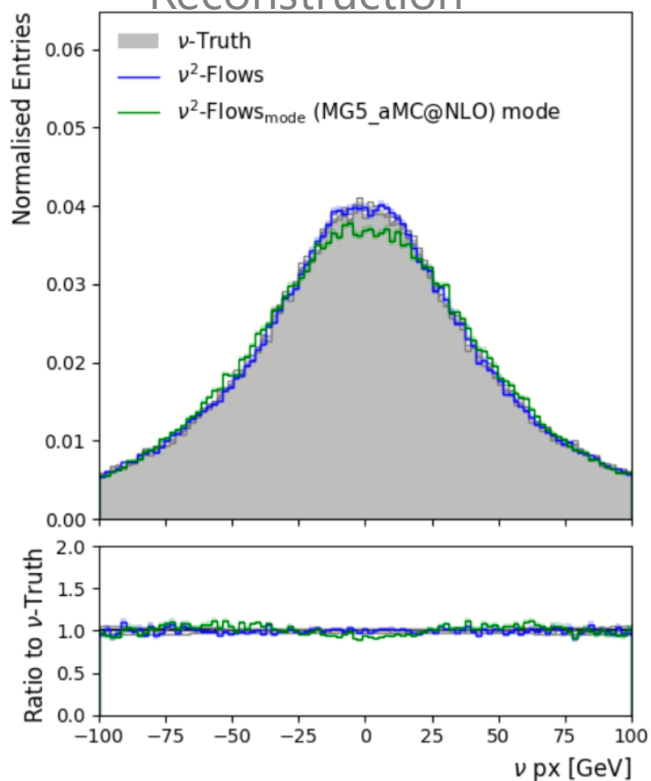
- nu2Flow: use **Normalization Flow** to generate the distribution of the neutrino



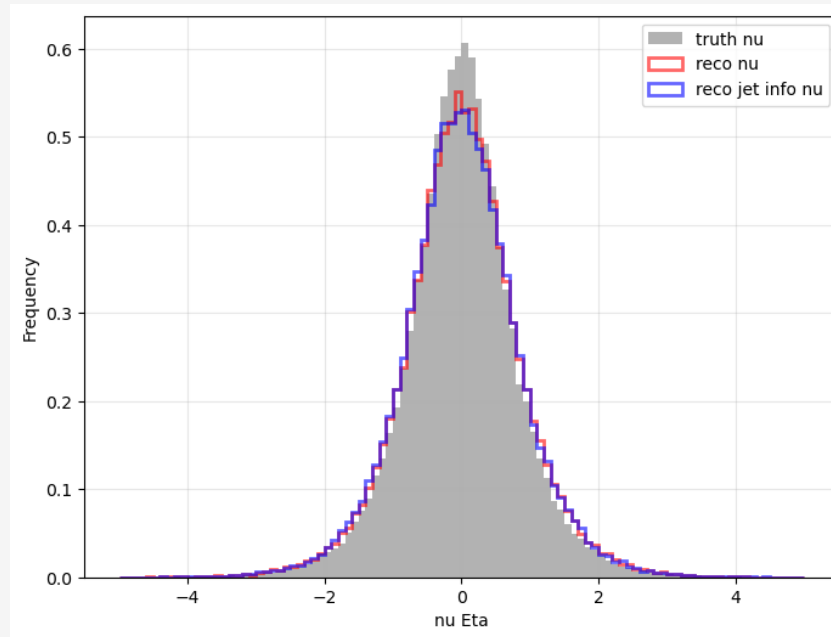
# SPA-Net Performance with Neutrino Reconstruction

- SPA-Net and Nu2Flow can model the neutrino's momentum well
- Using neutrino momentum information significantly improves top quark reconstruction efficiency
  - SPANet's neutrino regression improves  $t_h$  efficiency by 100% and  $t_l$  by 18%

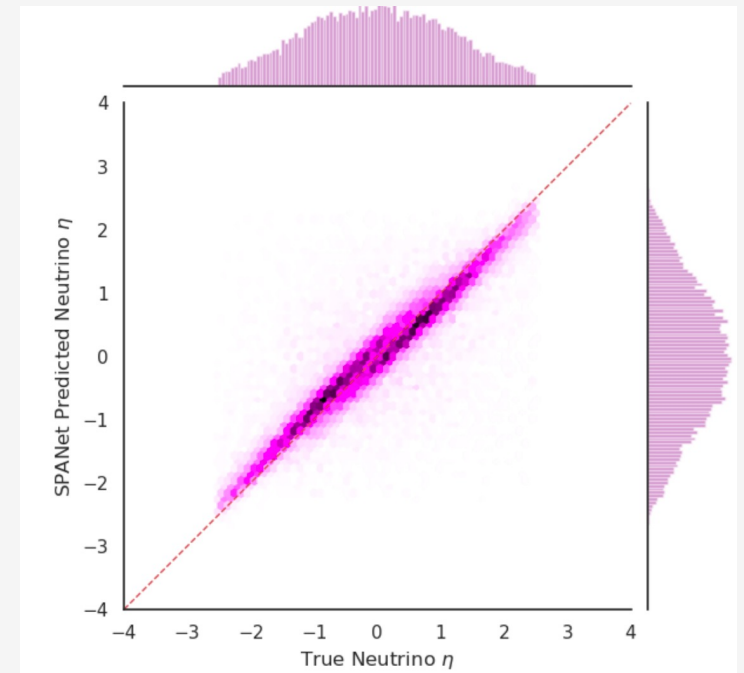
nu2Flow Neutrino  
Reconstruction



SPA-Net Neutrino Regression



SPA-Net Neutrino Regression



## Further Improvements to SPA-Net ?

- Losing the b jet requirement in top quark reconstruction: ~20% improvements in hadronic top
- Adding neutrino momentum from nu2Flow as the input of the SPA-Net
  - Further improves  $t_h$  efficiency by ~10%
  - Expect smaller improvement in  $t_l$  efficiency (requiring both b-quark and lepton)

SPA-Net+neutrino regression

Event Type	N jet	Hadronic top	Leptonic top
All Event	inclusive	0.383	0.703

nu2Flow + SPA-Net

Event Type	N jet	Hadronic top	Leptonic top
All Event	inclusive	0.415	0.628

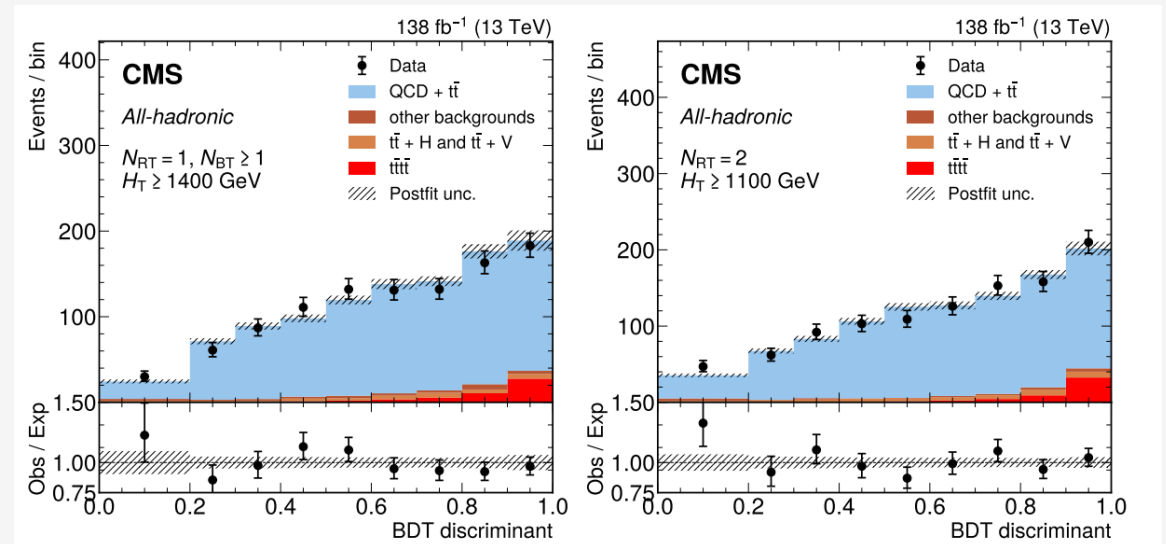
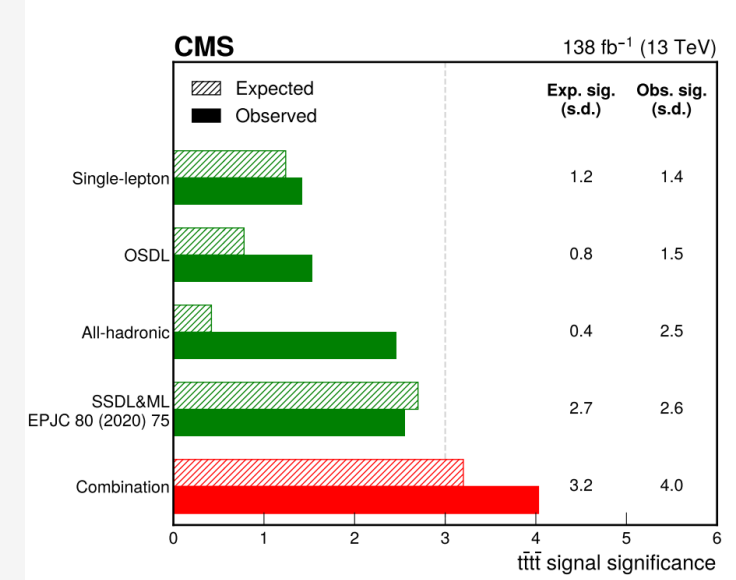
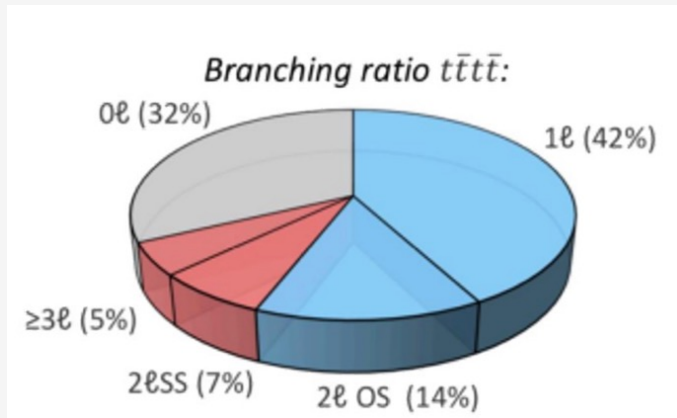
SPA-Net+no b-tagging requirement

Event Type	N jet	Hadronic top	Leptonic top
All Event	inclusive	0.502	0.701

Including large-R jets?  
Better b-tagging?  
Better MET/neutrino modeling?

# Four-top-quark Production in All-hadronic Channel

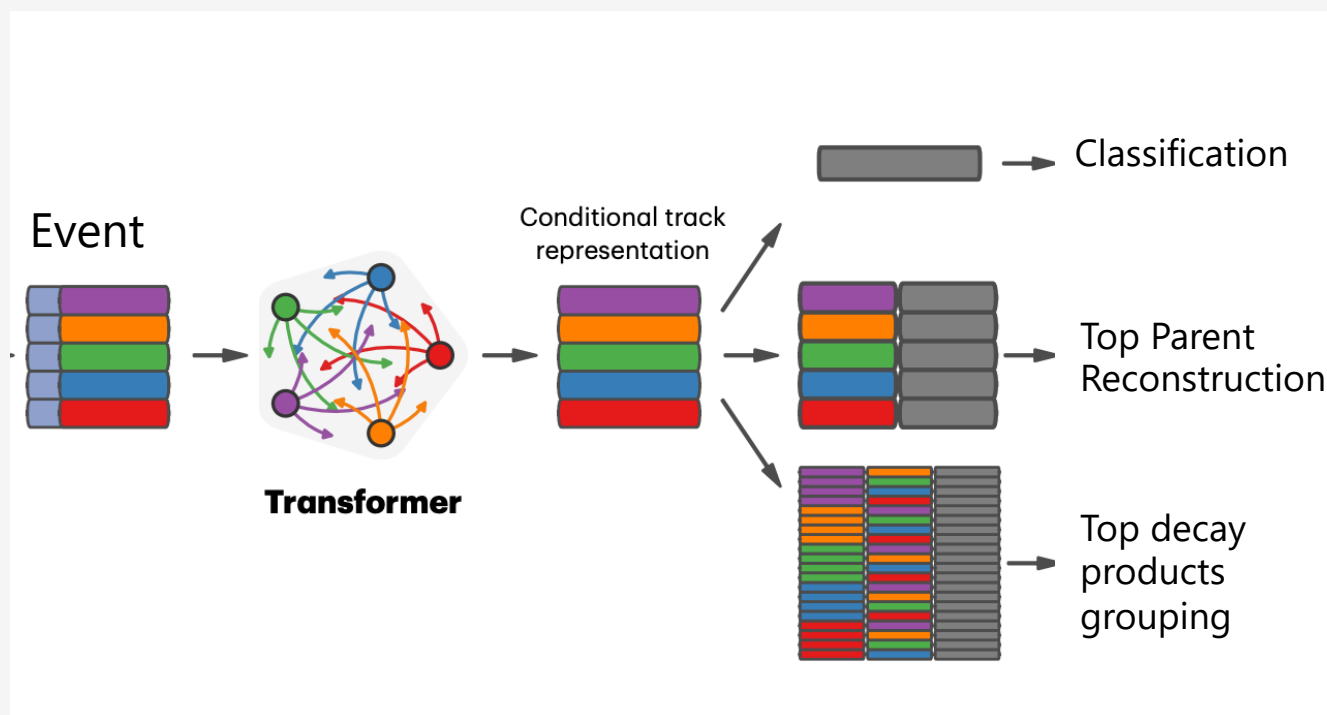
- Although SSML channel has better sensitivity due to less background but its decay branch ratio is 13%, 2 times smaller than the BR of all-hadronic channel
- CMS derived 2.7 sigma in the all-hadronic channel, using data-driven estimation for QCD background



# Transformer-based Framework for Four-top-quark Classification

Using the similar structure from the ATLAS transformer algorithm (GN2) but the input becomes the event level

- Considering dominated backgrounds, especially ttbar+jets
- Model inputs: jets kinematics and jet constituent-level information, event-level information



# Four-top-quark Reconstruction in Full hadronic Channel

- Adding the top quark reconstruction task in the event-level GN2 framework
- Utilizing the top quark parent ID as the mask, trying to do the reconstruction like the image masking

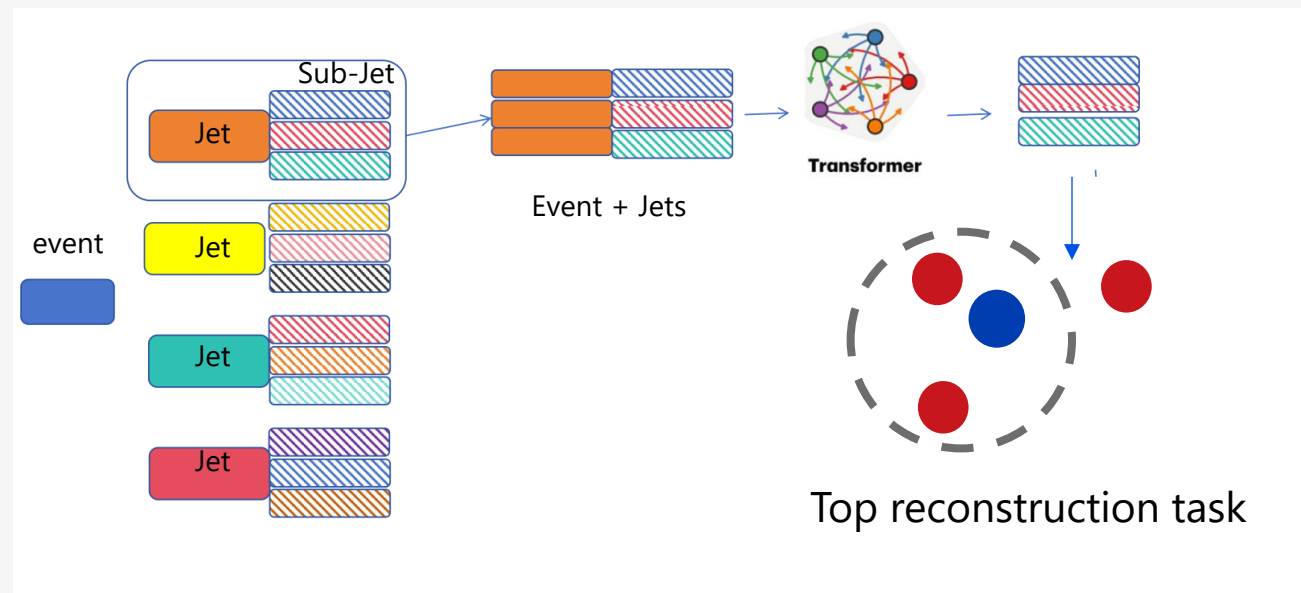
SPA-Net Algorithm

	$N_{\text{jets}}$	Event Fraction	SPA-NET Efficiency	
			Event	Top Quark
All Events	$\leq 12$	0.219	0.276	0.484
	$\leq 13$	0.304	0.247	0.474
	$\geq 14$	0.450	0.198	0.450
	<b>Inclusive</b>	<b>0.974</b>	<b>0.231</b>	<b>0.464</b>

SciPost Phys. 12, 178 (2022)

- SPA-Net reaches 46% reconstruction efficiency in all-hadronic channel
- Long training time due to large number of training parameters (16M)

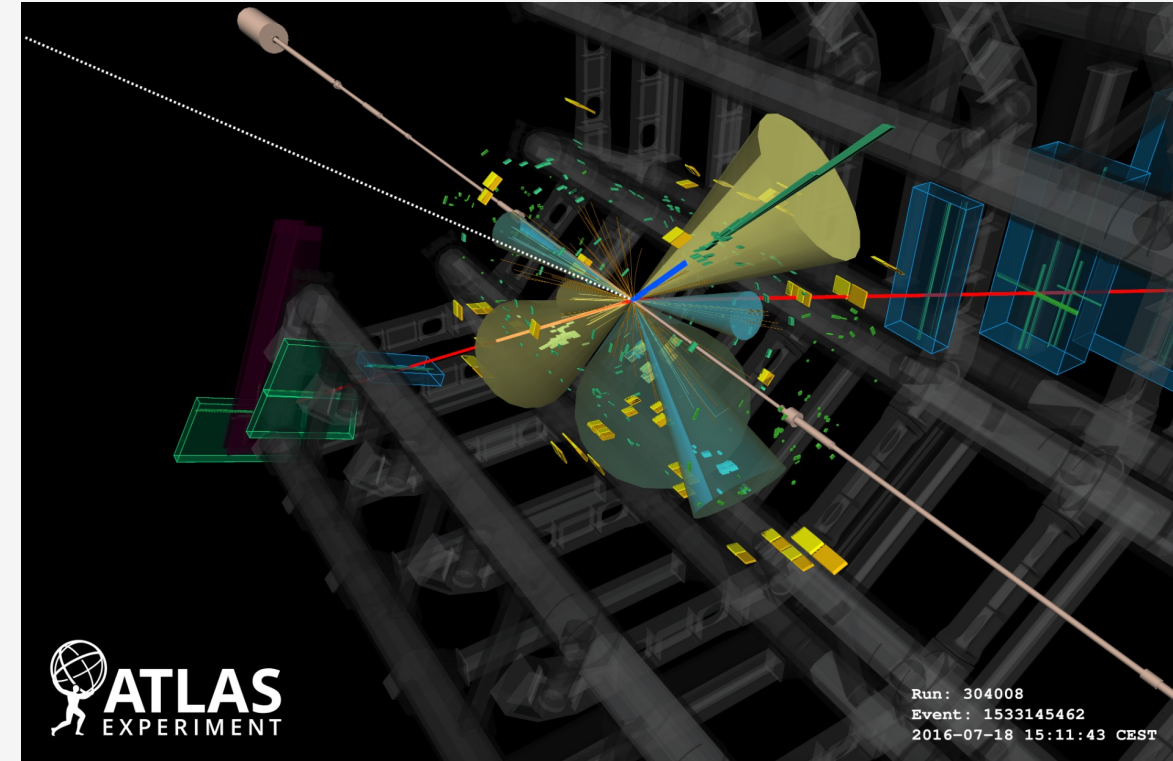
GN2 Framework



Study on-going

# Summary and Outlook

- Top quark reconstruction can reduce the multi-top background in the analysis
- Machine learning method can be applied for the further top reconstruction in the different 4top final state
  - Using SPA-Net for top reconstruction in SSML channel: **3 times** faster and hadronic top efficiency doubled to be **~40%**, leptonic top efficiency reaches **70%**
  - Further optimizations such as neutrino reconstruction and jet substructures
- The sensitivity in the full-hadronic channel can be enhanced by the machine learning method
  - Classification and top quark reconstruction task using the event-level GN2 algorithm





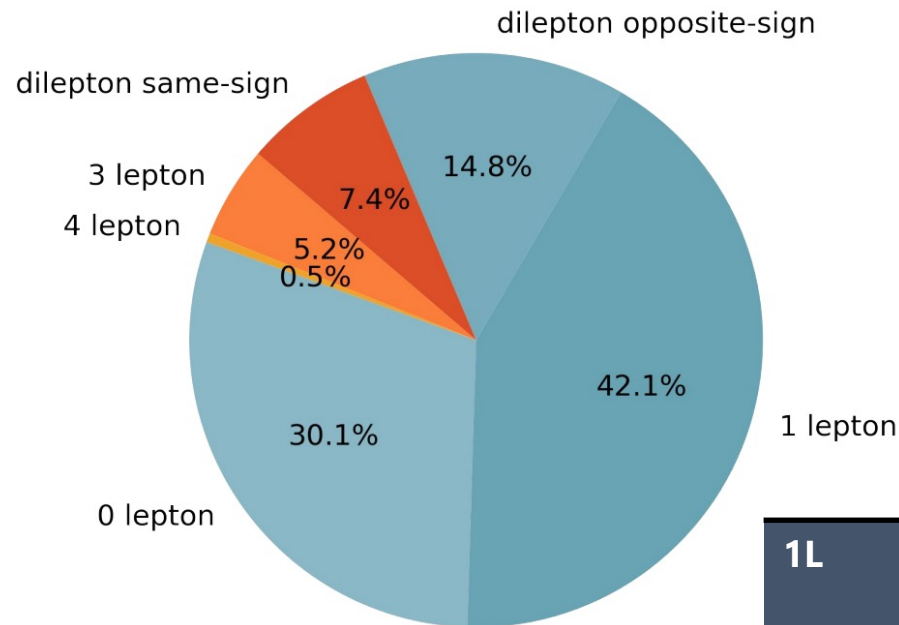


# Backups

# SPA-Net Performance in Different 4top channels

- Test the SPA-Net in different 4top channels
- 2L channel has the best performance overall in top quark reconstruction efficiency
- With the less hadronic top quarks, better performance

All-had	Hadronic top
Efficiency	0.464



2L	Hadronic top	Leptonic top
Efficiency	0.383	0.703

1L	Hadronic top	Leptonic top
Efficiency	0.305	0.604