

Machine-Learning quantum entanglement with top quark pair production at the LHC

Zhongtian Dong

University of Kansas

Based on ongoing work in collaboration with

A. Serratos, D. Gonçalves, K.C. Kong,

Why top quark pair production?

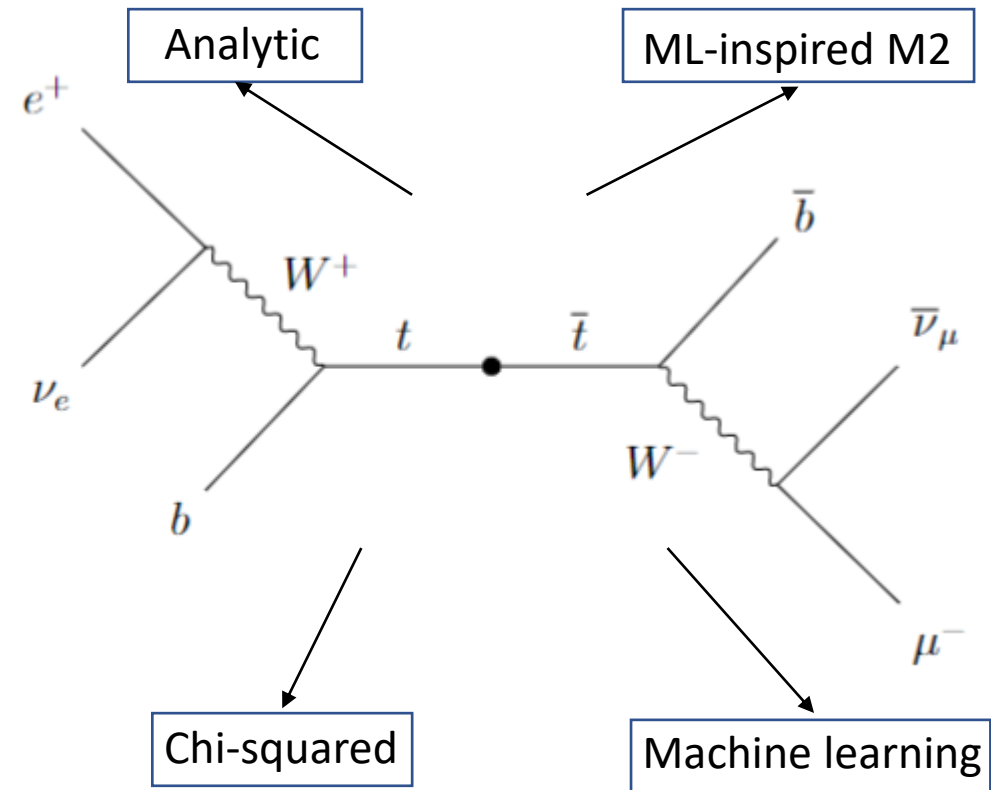
- Growing interest in observing entanglement and violation of Bell's inequality at the LHC. 2106.01377 Barr 2102.11883 Fabbrichesi, Floreanini, Panizzo 2110.10112 Severi, Boschi, Maltoni, Sioli
- Top quark pair productions are good candidate because they decay before de-correlation and that the lepton from two-step decay is highly correlated with Top spin.

$$\frac{1}{\Gamma} \frac{d\Gamma}{d \cos \chi} = \frac{1 + \alpha \cos \chi}{2} = \begin{cases} +1.0 & l^+ \text{ or } \bar{d} \\ -0.31 & \nu \text{ or } u \\ -0.41 & b \end{cases} \quad 1001.3422 \text{ Mahlon, Parke}$$

- χ is the angle between the decay product and top quark spin axis in the top quark rest frame.
- To obtain optimal basis for spin-correlation study, we need to reconstruct top quarks in dileptonic $t\bar{t}$ events.

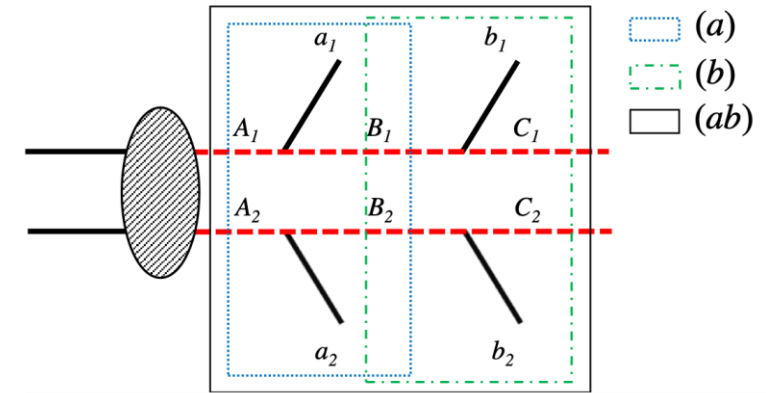
Top quark reconstruction

- To maximally exclude backgrounds, we study events with different flavored lepton final states.
- We need to find the neutrino momentum and the correct b quark and lepton pairing given all the visible momentum.
- We compared the performance of different reconstruction methods cross different statistics and chose one to do analysis.
- Event generation using MG5 for event generation. PYTHIA8 for parton-shower and hadronization. Delphes for detector effect



ML-inspired M2 reconstruction

- Generalization of MT2, which allows additional mass constraints.
- Obtain neutrino momentum from minimization.
- Reconstructed mass distribution of the non-constrained mass is not optimal.
- Accuracy of combinatorial assignment is better when using neural network, we used it to resolve combinatorics instead of M2.



$$M_{2CW}^{(b\ell)} \equiv \min_{\vec{q}_1, \vec{q}_2} \{ \max [M_{P_1}(\vec{q}_1, \tilde{m}), M_{P_2}(\vec{q}_2, \tilde{m})] \}$$

$$\vec{q}_{1T} + \vec{q}_{2T} = \vec{\cancel{p}}_T$$

$$M_{P_1} = M_{P_2}$$

$$M_{R_1}^2 = M_{R_2}^2 = m_W^2$$

$$M_{2Ct}^{(\ell)} \equiv \min_{\vec{q}_1, \vec{q}_2} \{ \max [M_{P_1}(\vec{q}_1, \tilde{m}), M_{P_2}(\vec{q}_2, \tilde{m})] \}$$

$$\vec{q}_{1T} + \vec{q}_{2T} = \vec{\cancel{p}}_T$$

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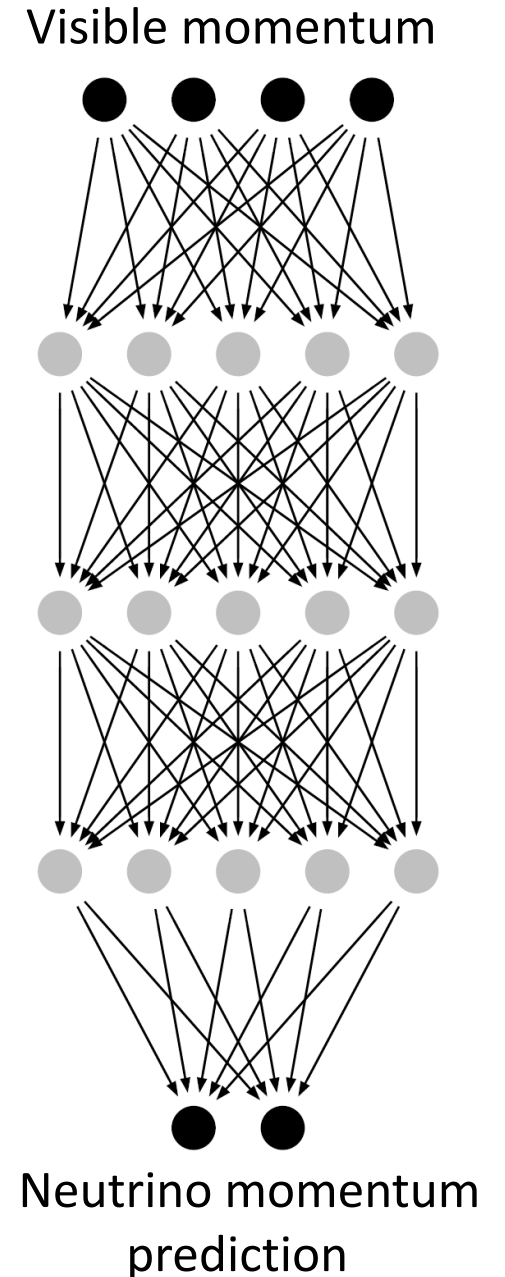
$$M_{R_1}^2 = M_{R_2}^2 = m_t^2$$

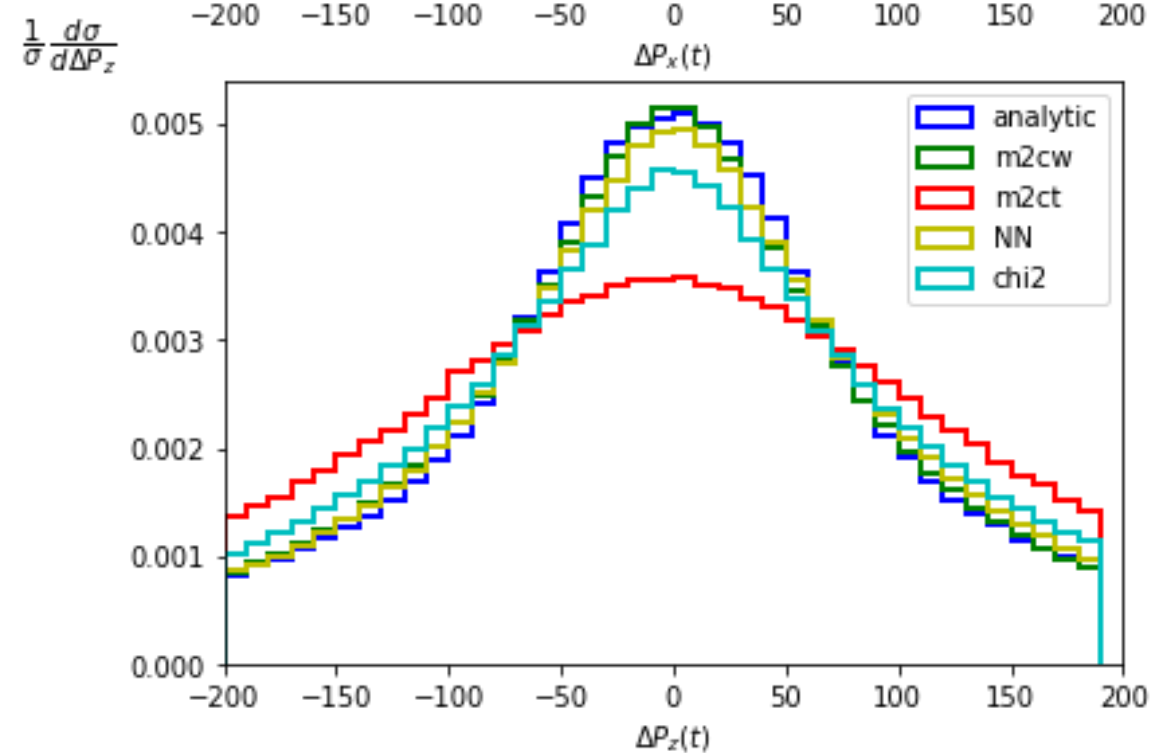
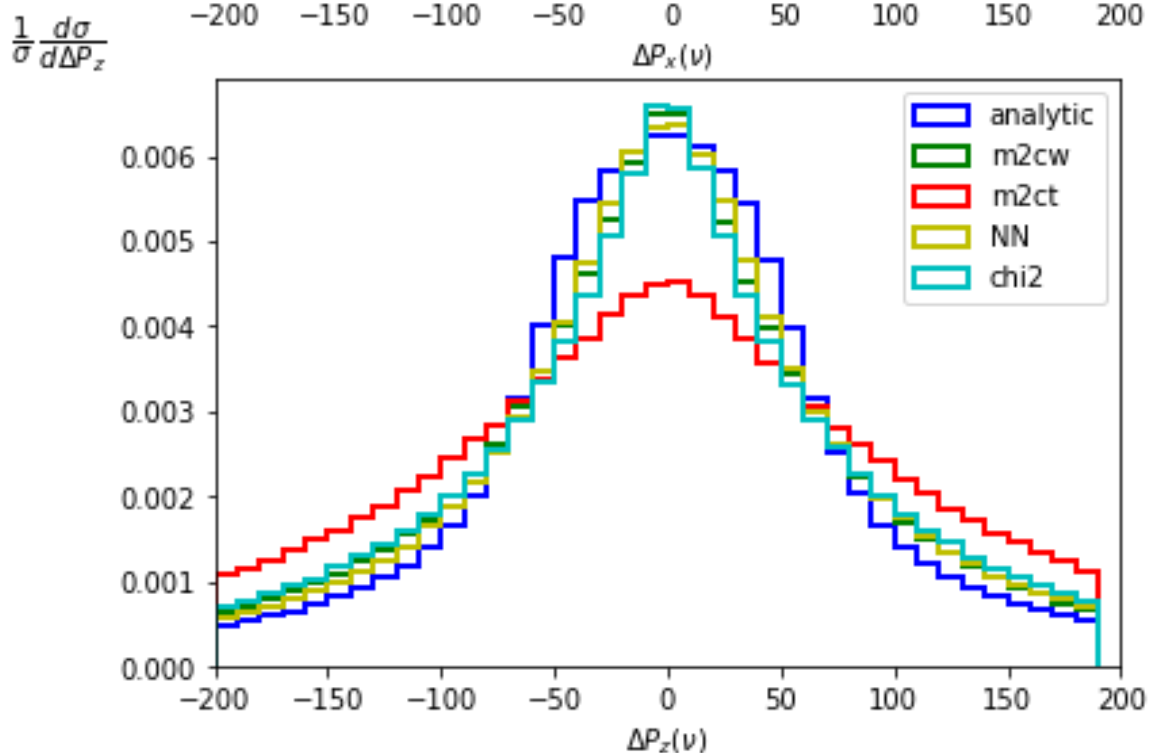
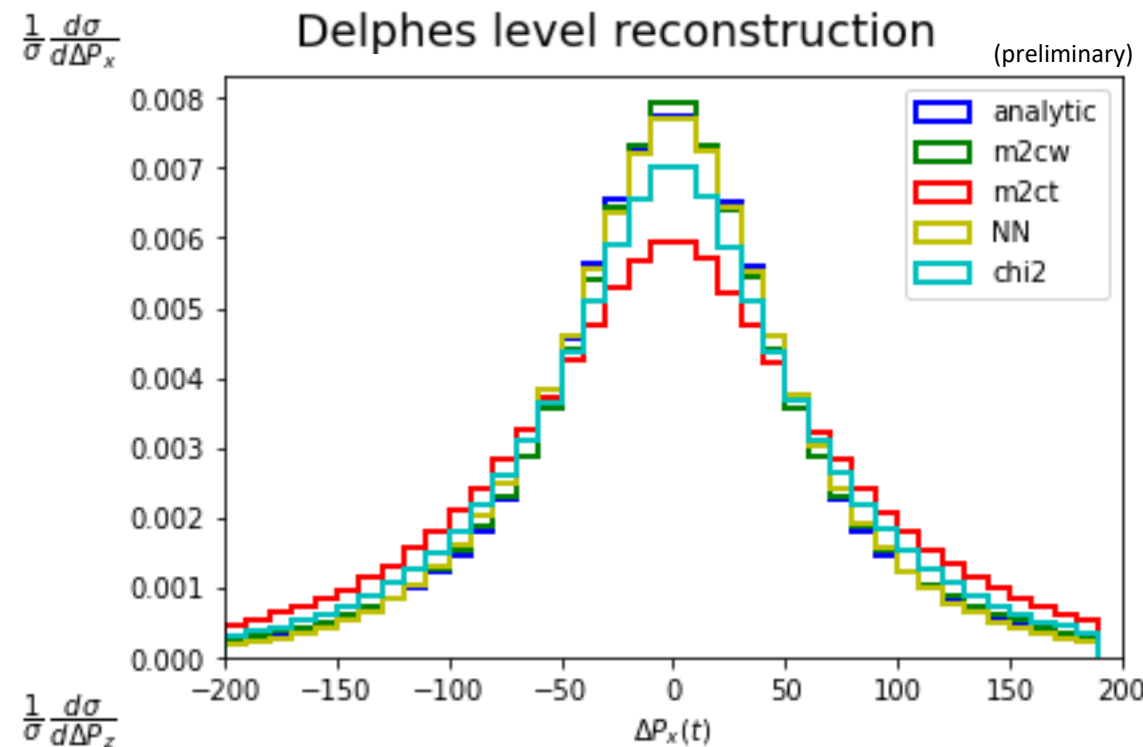
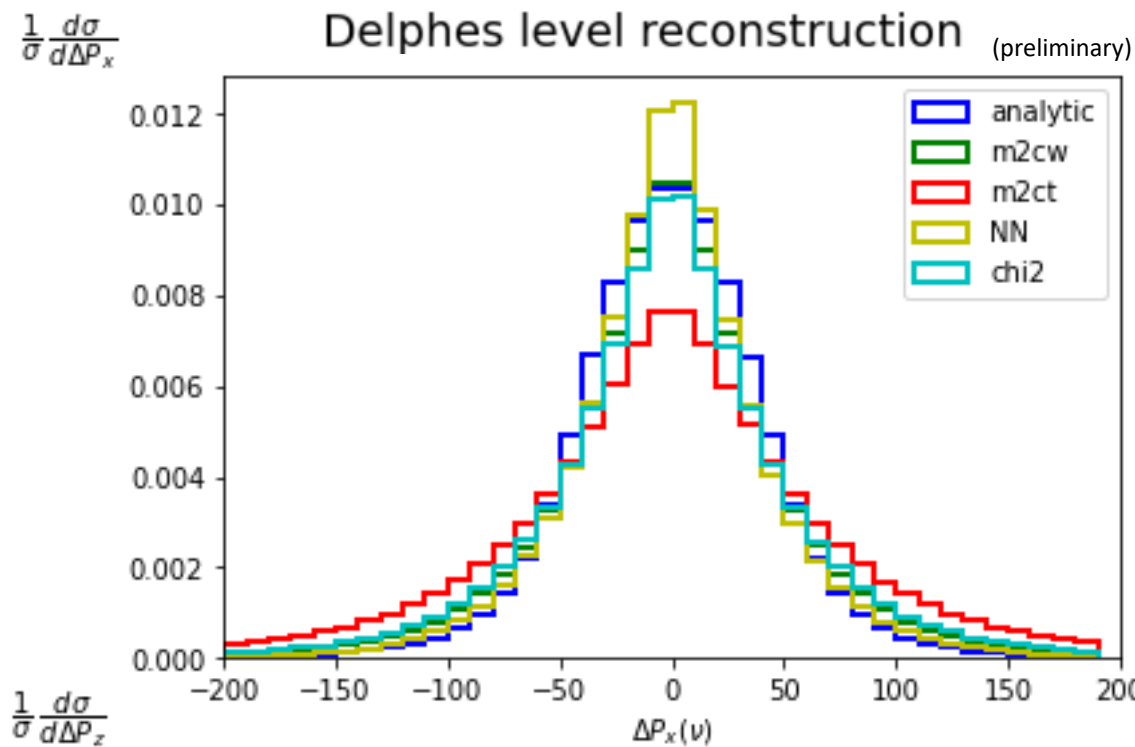
Machine learning reconstruction

- Fully connected DNN trained on parton level data.
- Uses mean squared loss of neutrino momentum.
- Adds additional mean squared loss of Top and W masses, as inspired from the chi-squared method.

$$L = \sum_{i=1}^6 (\hat{p}_i - p_i)^2 + \lambda_1 [(\hat{m}_t - m_t)^2 + (\hat{m}_{\bar{t}} - m_t)^2] + \lambda_2 [(\hat{m}_{W^+} - m_W)^2 + (\hat{m}_{W^-} - m_W)^2]$$

- A separate network resolves combinatorial assignment of b quark and leptons.

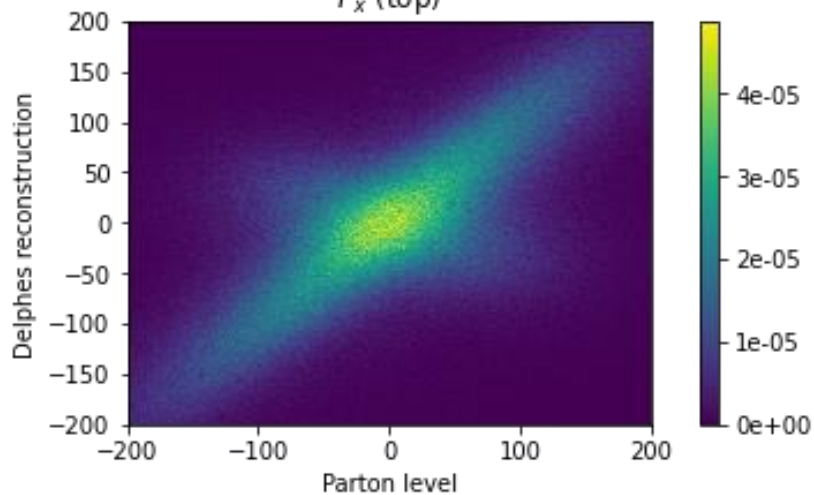




Analytic

(preliminary)

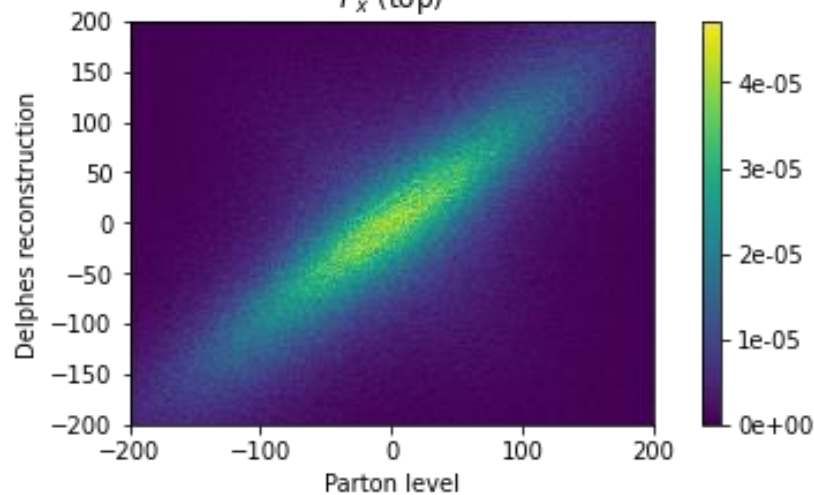
P_x (top)



ML-inspired M2

(preliminary)

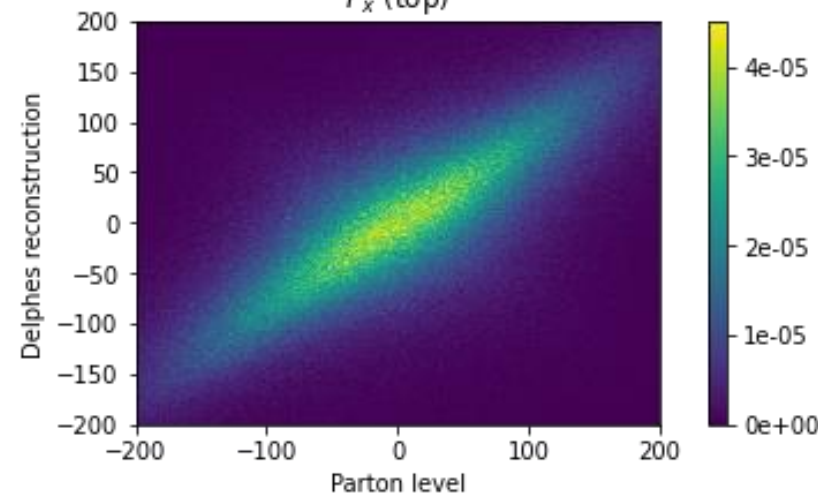
P_x (top)



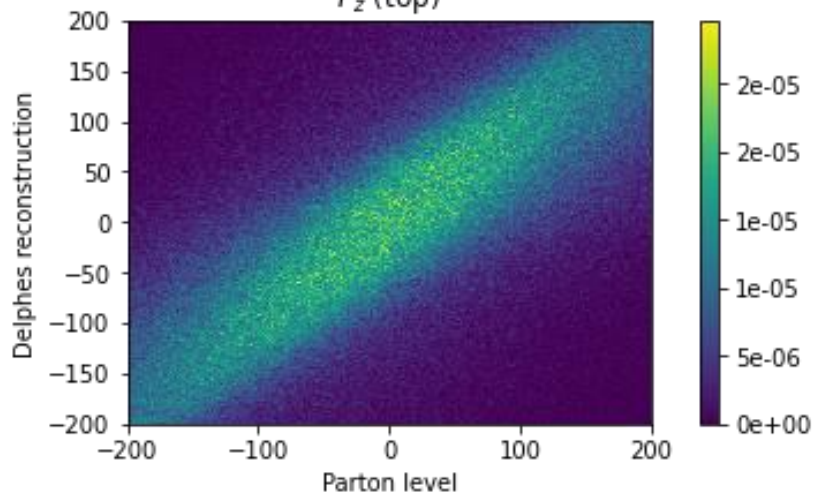
Neural Network

(preliminary)

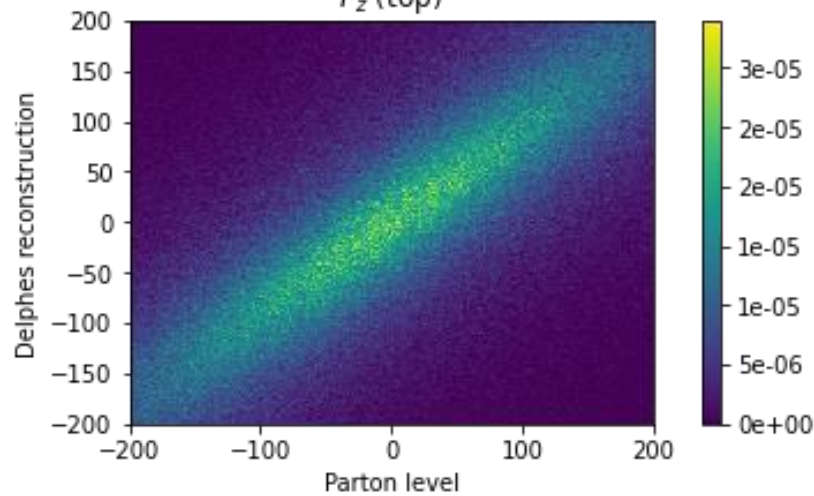
P_x (top)



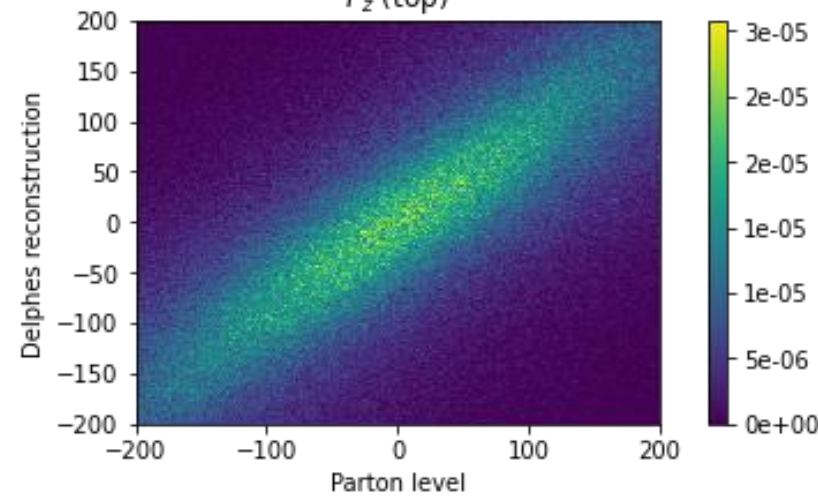
P_z (top)



P_z (top)



P_z (top)



Top quark production as a two-qubit system

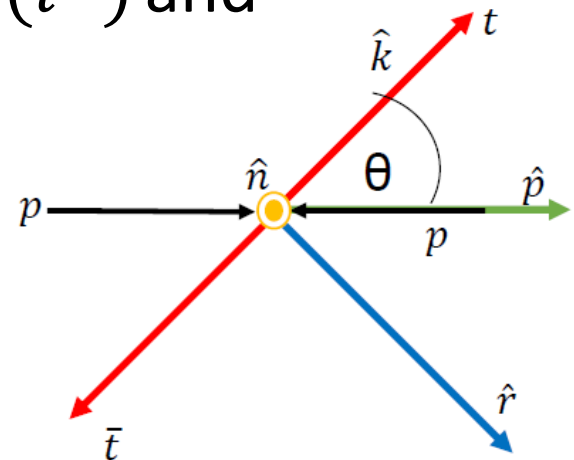
For a system of two spin-1/2 particles, the density matrix can be written as

$$\rho = \frac{I_4 + B_i^+ \sigma^i \otimes I_2 + B_i^- I_2 \otimes \sigma^i + C_{ij} \sigma^i \otimes \sigma^j}{4}$$

$C_{ij} = \langle \sigma^i \otimes \sigma^j \rangle$ Represents the spin correlation between the two particles.

We can compute the C_{ij} by computing angles between l^+ (l^-) and the axis in t (\bar{t}) rest frames respectively.

$$\hat{k} = \text{top direction}, \quad \hat{r} = \frac{\hat{p} - \hat{k} \cos \theta}{\sin \theta}, \quad \hat{n} = \hat{k} \times \hat{r}, \quad \hat{p} = (0, 0, 1)$$



Entanglement

For a system of two spin-1/2 particles, the criterion for entanglement is $|C_{11} + C_{22}| - C_{33} > 1$, equivalent under permutation.

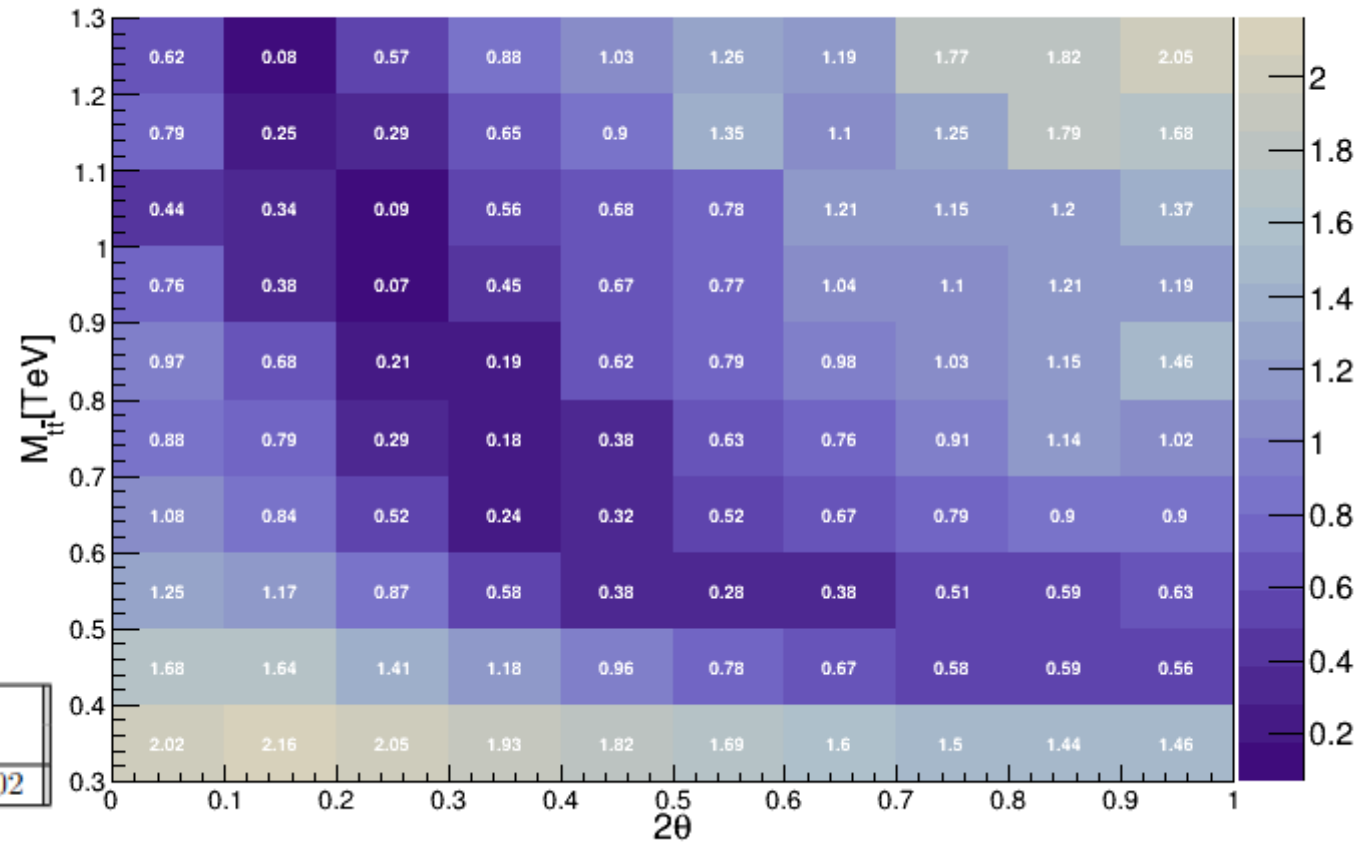
2203.05582, Y. Afik, J. R. M. de Nova
2110.10112, C. Severi et al.

For the basis we defined, we only need to check

$$|C_{kk} + C_{rr}| - C_{nn}$$

Region	$ C_{kk} + C_{rr} - C_{nn}$ (Parton)
Threshold	1.60 ± 0.02

$$-C_{nn} + |C_{kk} + C_{rr}| \quad \text{(preliminary)}$$



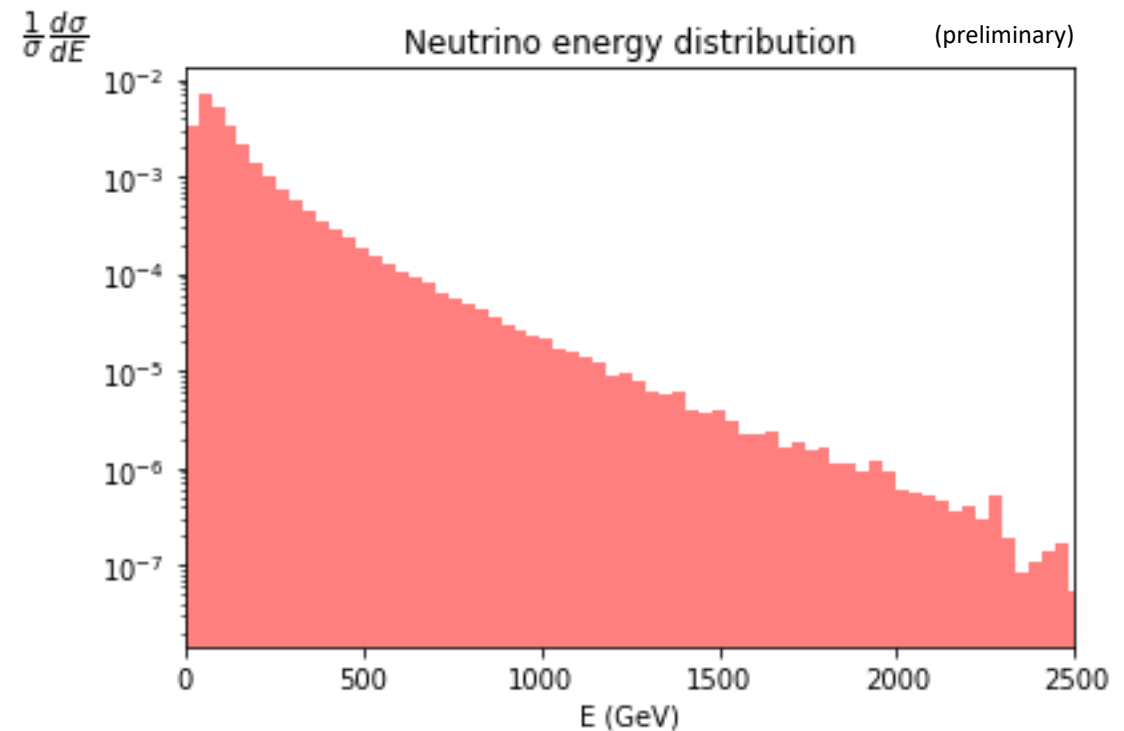
Region	$-C_{nn} + C_{kk} + C_{rr} $ (Delphes)				NN
	Analytic	ML-M2CW	ML-M2CT	χ^2	
Threshold	0.73 ± 0.02	1.33 ± 0.02	3.71 ± 0.02	1.00 ± 0.02	2.02 ± 0.02

Summary

- Top quark pair-production at the LHC provides a window to study the foundations of QM in the high-energy regime.
- The dilepton final state is useful as the top (anti-top) and lepton spins are fully correlated.
- We should choose the reconstruction method that best fits our purpose.
- The ML-M2CW method performs better than others in the reconstruction of the entanglement signature.
- We should continue to explore semi-leptonic channel.

Analytic reconstruction

- Based on the four on-shell conditions.
- Involves solving quartic polynomial with potential multiple real solutions.
- We compute weighted solution based on neutrino energy distribution.
- Does not consider widths of Top and W masses. Reconstructed mass distribution is not optimal.



Entanglement Calculation

- We can either calculate the spin correlation matrix by estimating the differential cross-section or using anti-symmetry.

$$\frac{1}{4}\alpha_a\alpha_b = \frac{N(\cos\theta_a^i \cos\theta_b^i > 0) - N(\cos\theta_a^i \cos\theta_b^i < 0)}{N(\cos\theta_a^i \cos\theta_b^i > 0) + N(\cos\theta_a^i \cos\theta_b^i < 0)}$$

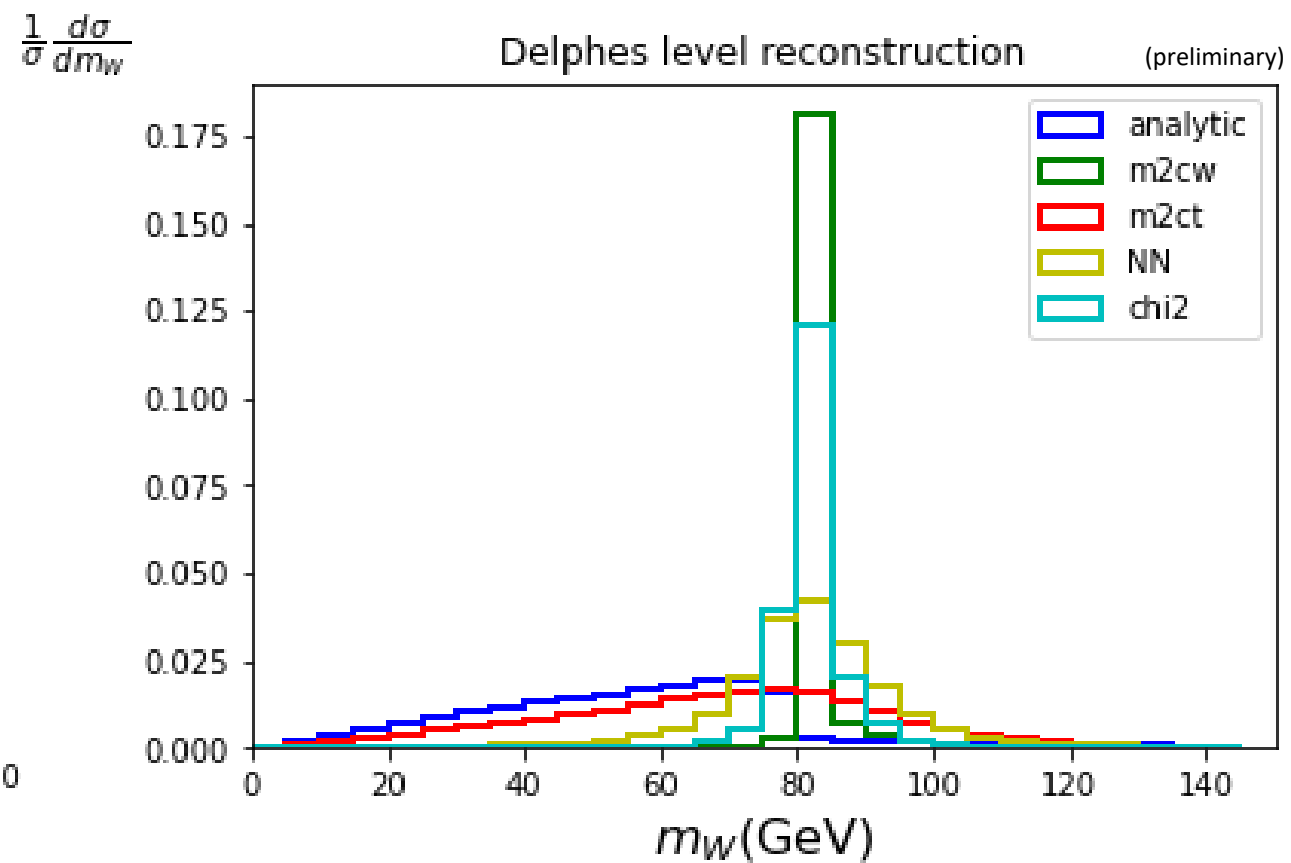
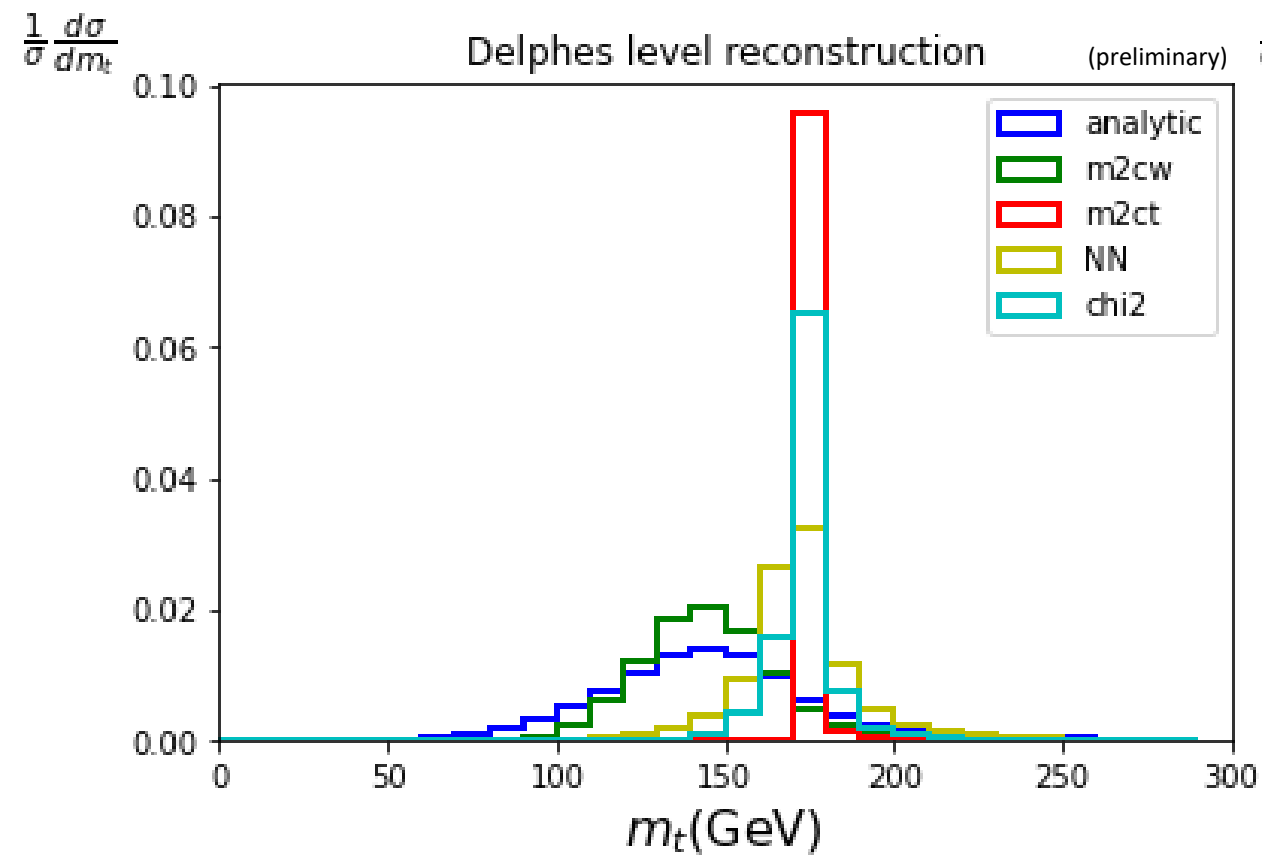
Chi-squared reconstruction

- Minimized over chi-squared statistic.
- Uses explicit mass information

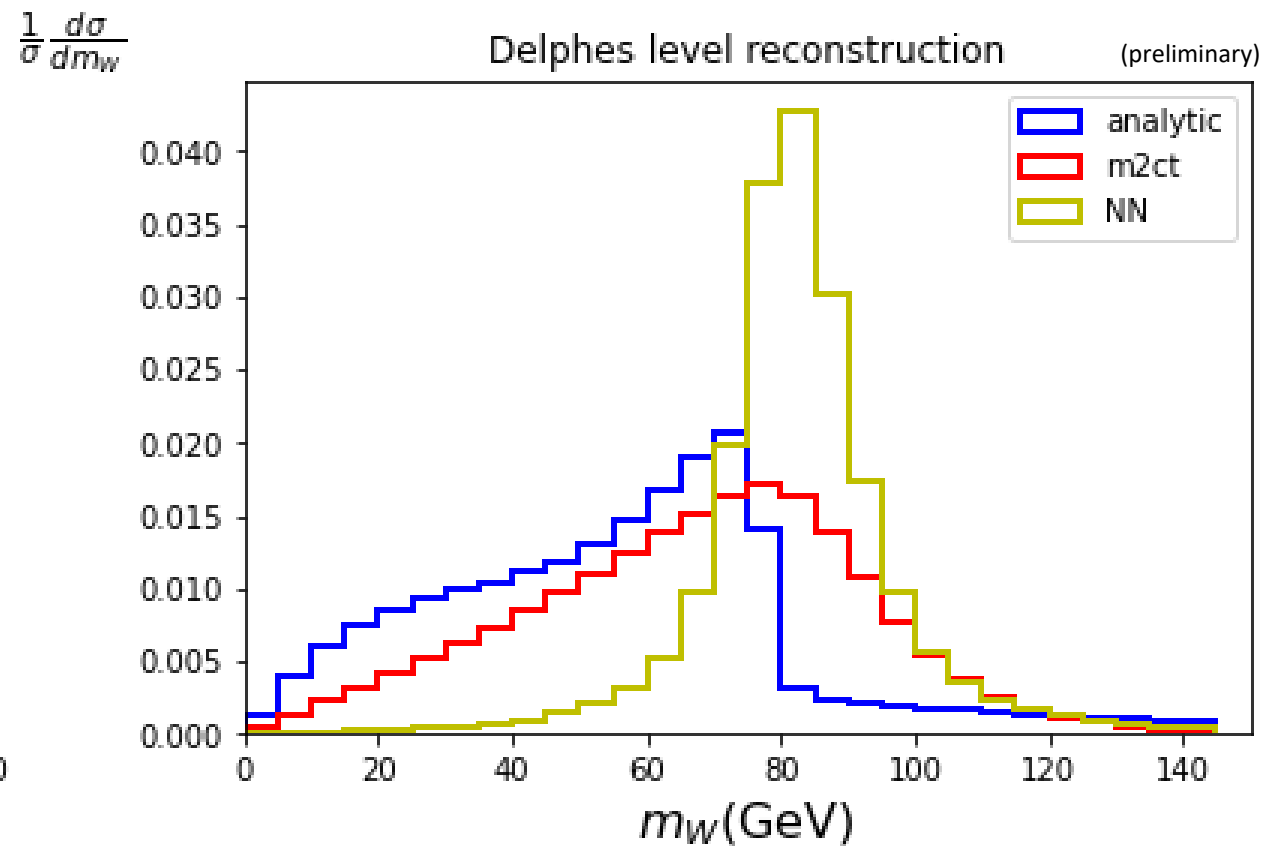
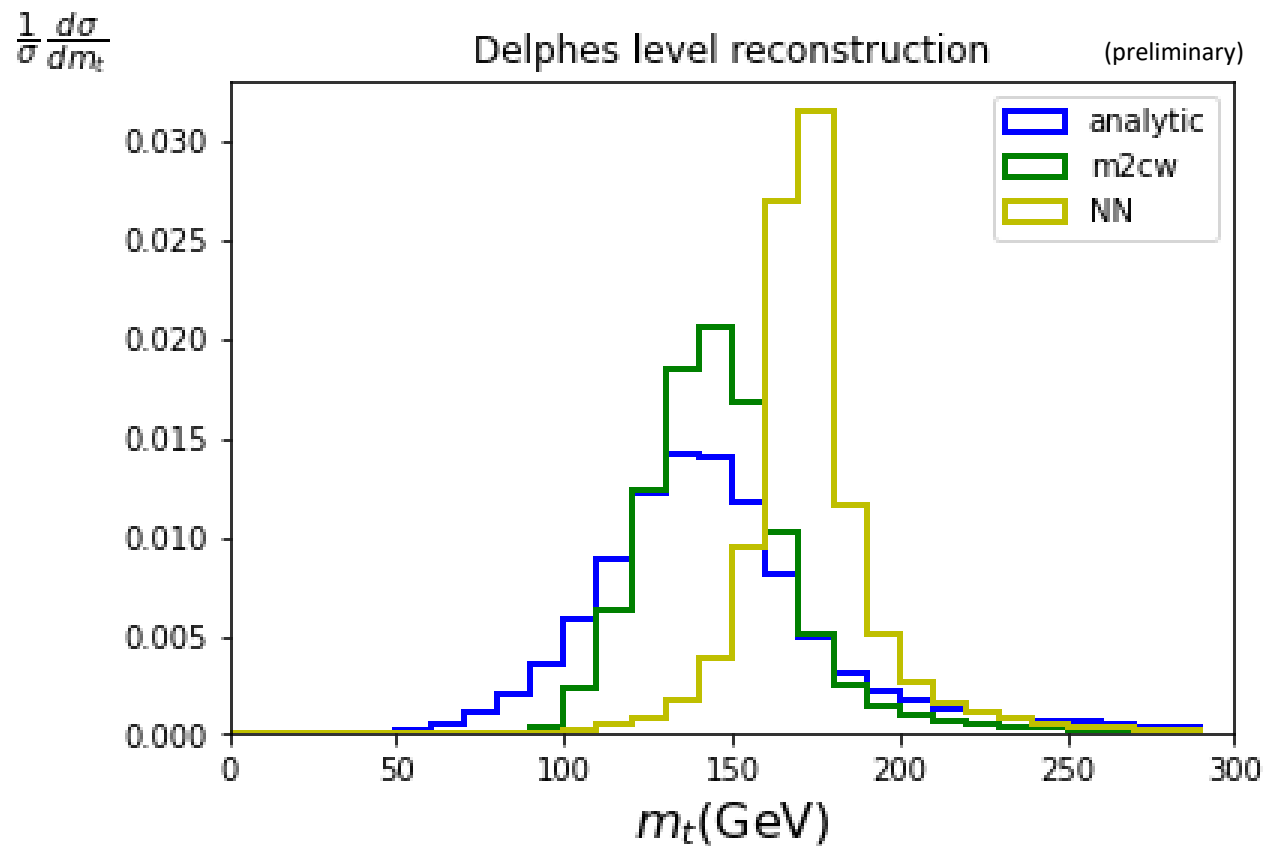
$$\chi^2 \equiv \min_{\not{p}_T = p_{\nu T} + p_{\bar{\nu} T}} \left[\frac{(m_{b\ell+\nu} - m_t)^2}{\sigma_t^2} + \frac{(m_{\ell+\nu} - m_W)^2}{\sigma_W^2} + \frac{(m_{b\bar{\ell}-\bar{\nu}} - m_t)^2}{\sigma_t^2} + \frac{(m_{\ell-\bar{\nu}} - m_W)^2}{\sigma_W^2} + \frac{(|p_{T_t}| - |p_{T_{\bar{t}}}|)^2}{\sigma_{pT}^2} \right]$$

- Resolves combinatorics by choosing the combination with a smaller chi-squared value.
- More hyperparameters are used compared to other methods.
- Cannot generalize to other mass configurations.

Comparisons



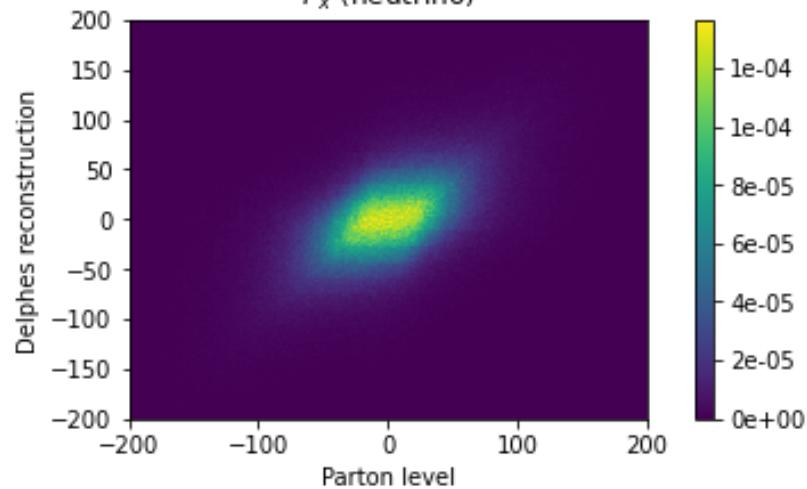
Comparisons



Analytic

P_x (neutrino)

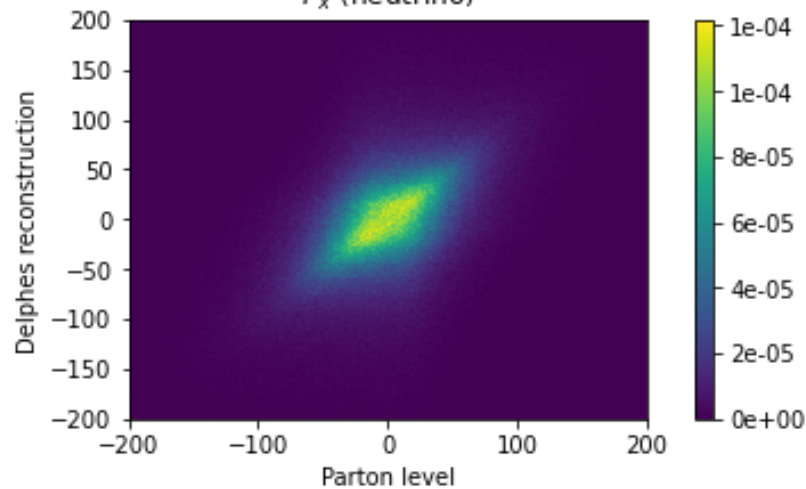
(preliminary)



ML-inspired M2

P_x (neutrino)

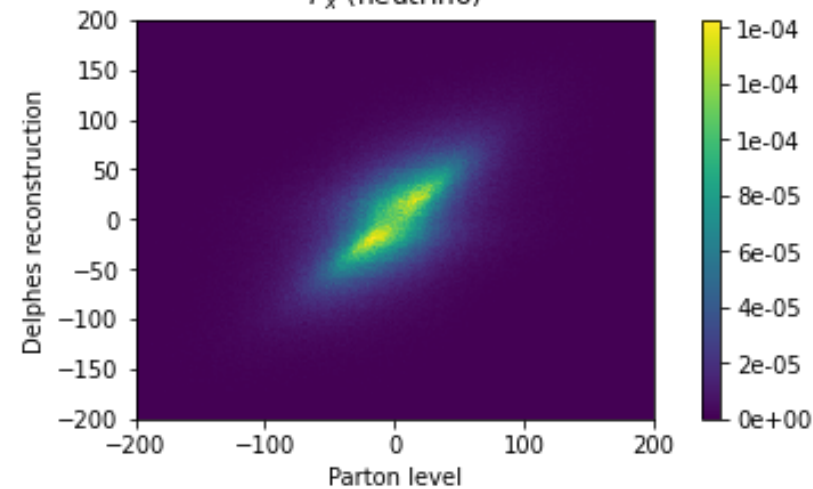
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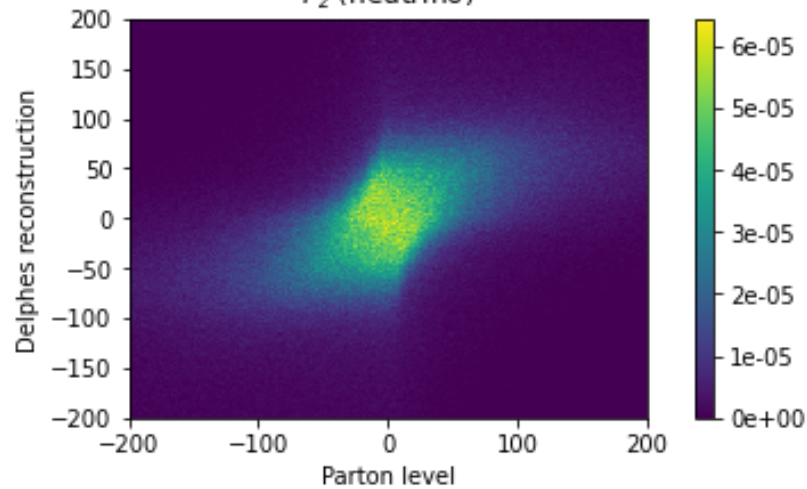
Neural Network

P_x (neutrino)

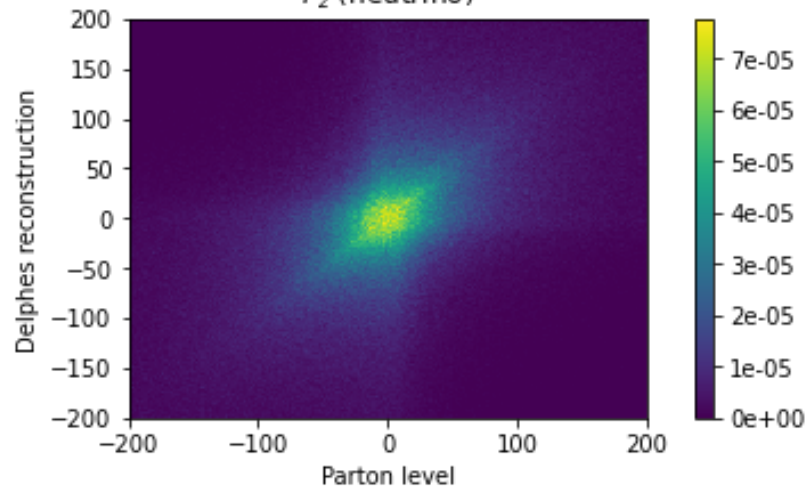
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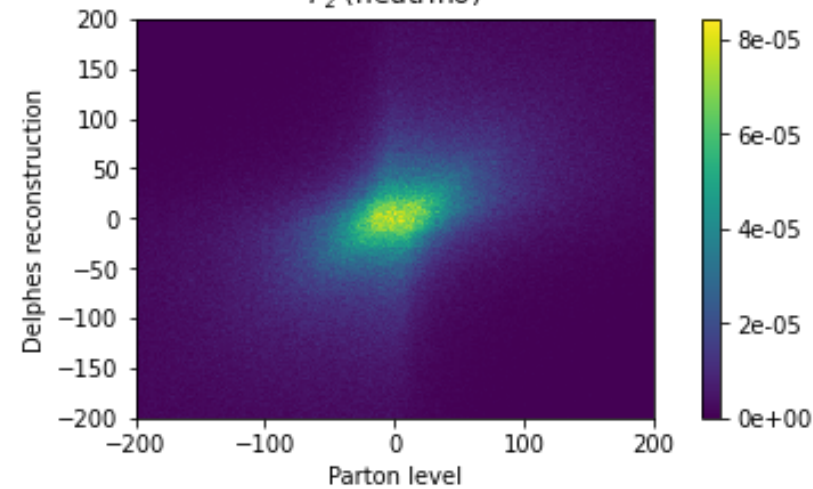
P_z (neutrino)



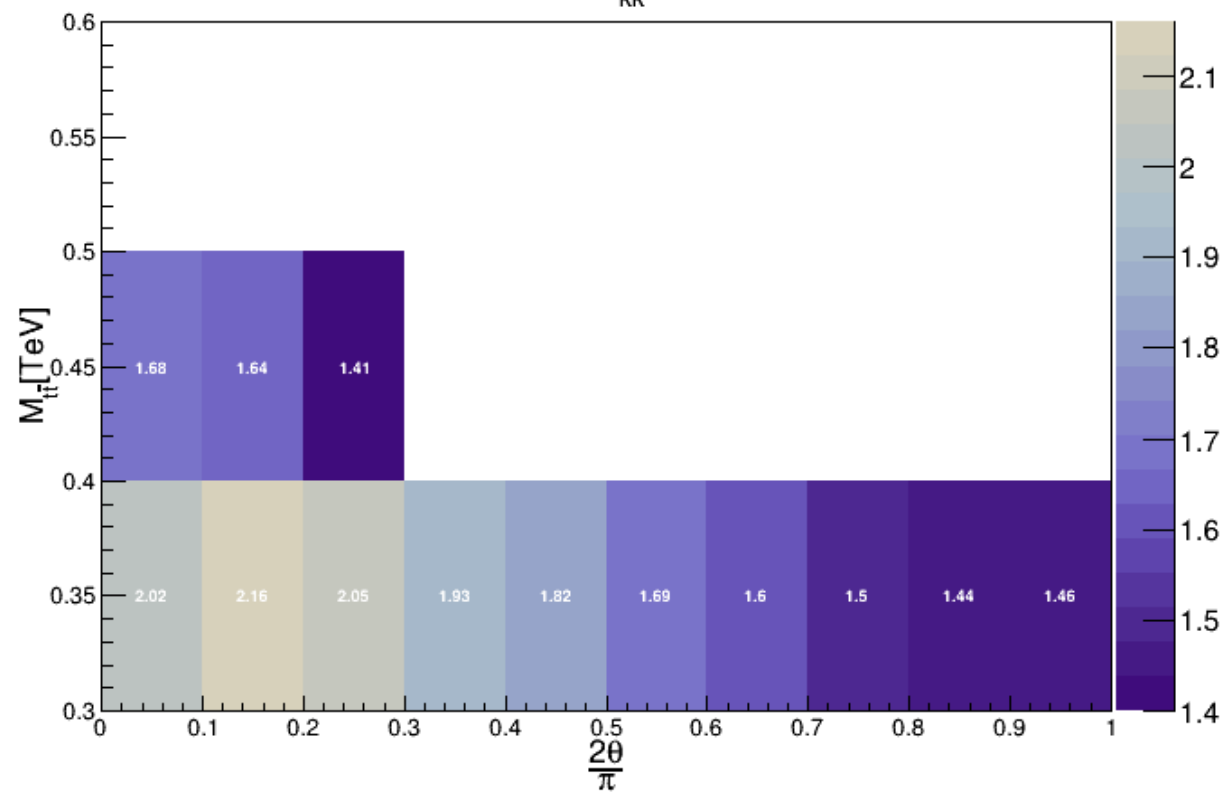
P_z (neutrino)



P_z (neutrino)



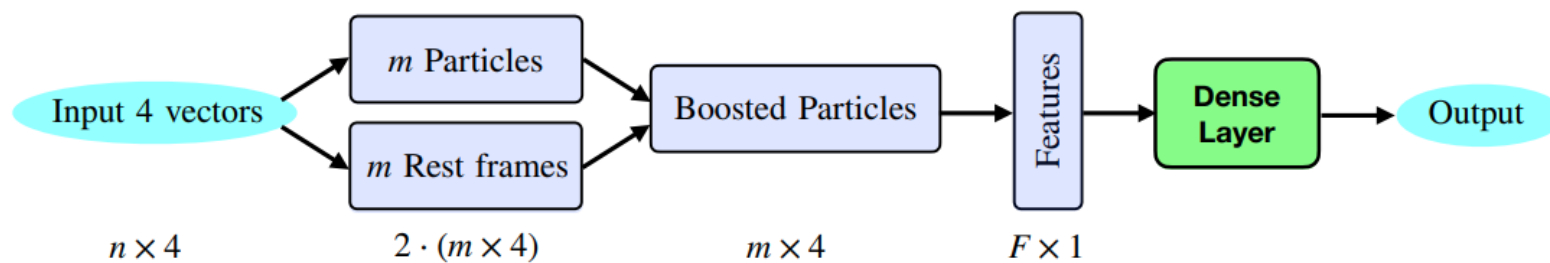
$$-C_{nn} + |C_{kk} + C_{rr}|$$



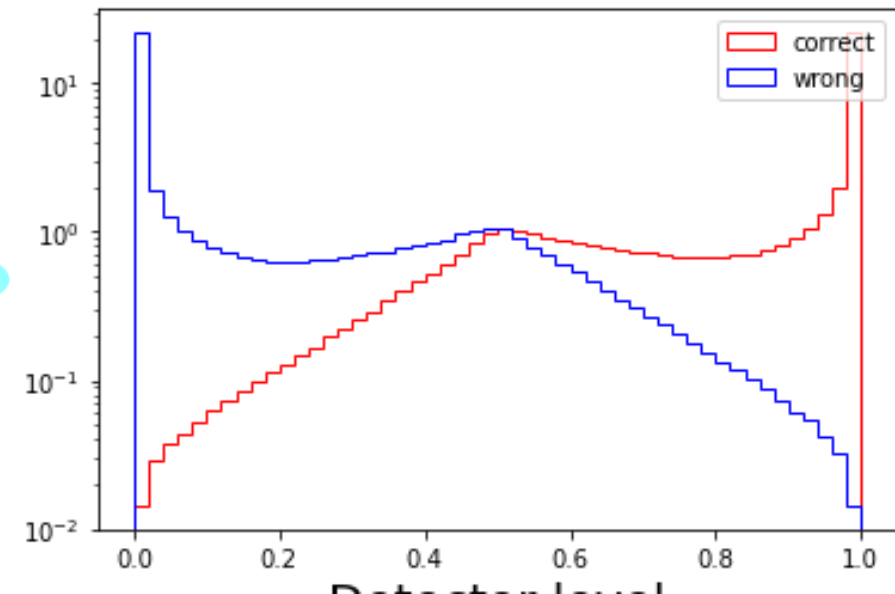
Combinatorial Classification: Lorentz Boosted Network

- Motivated by particle kinematics in the rest frames of various particle combinations.
- Form linear combinations of input momentums and boosts into rest frames.
- Output features of the boosted particles.
- We used the network structure studied in our previous work.

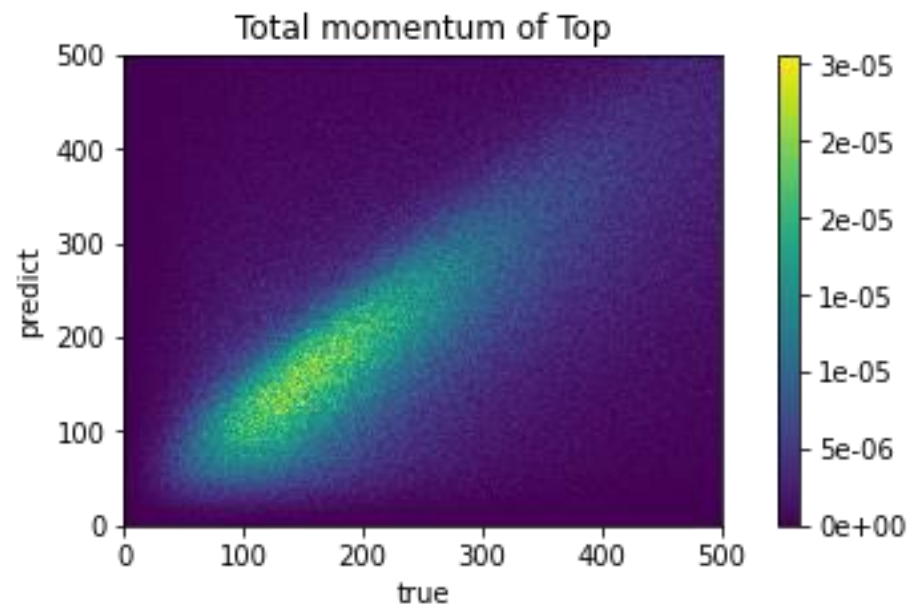
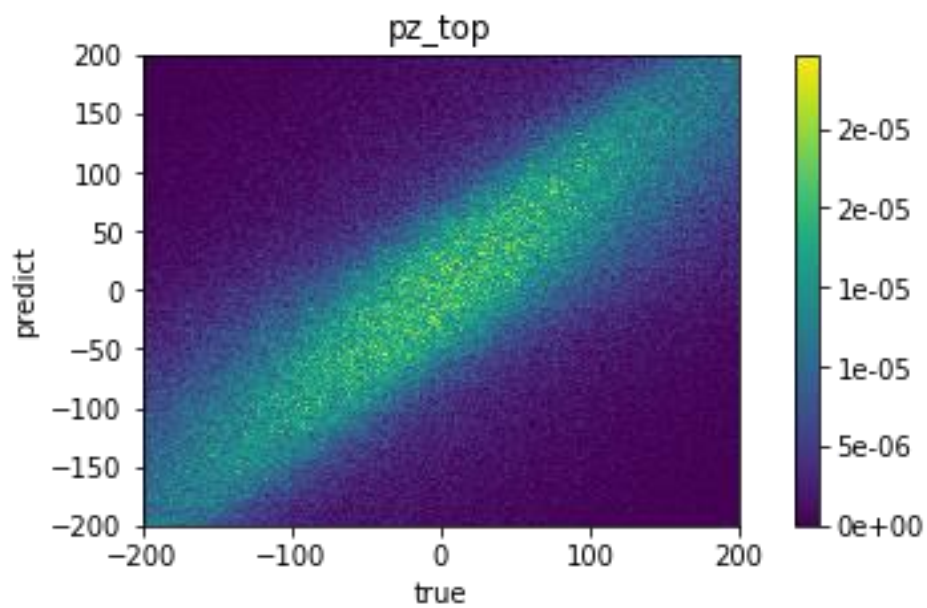
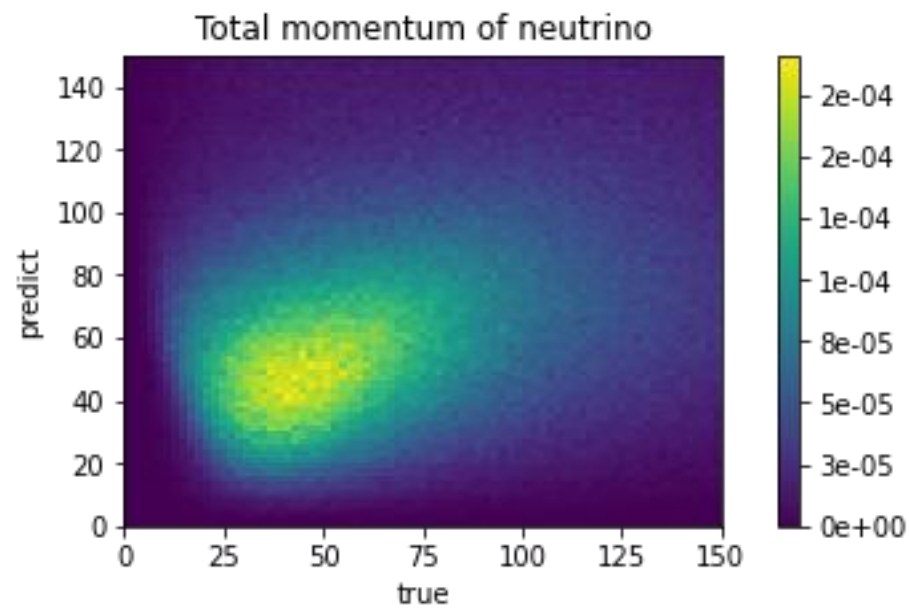
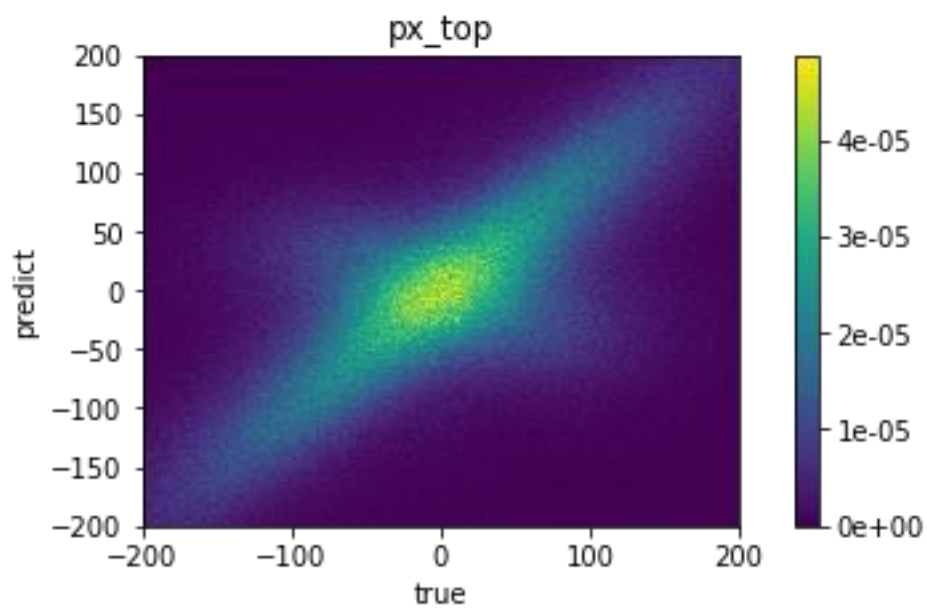
2202.05849 Alhazmi, Dong, Huang, Kim, Kong, Shih



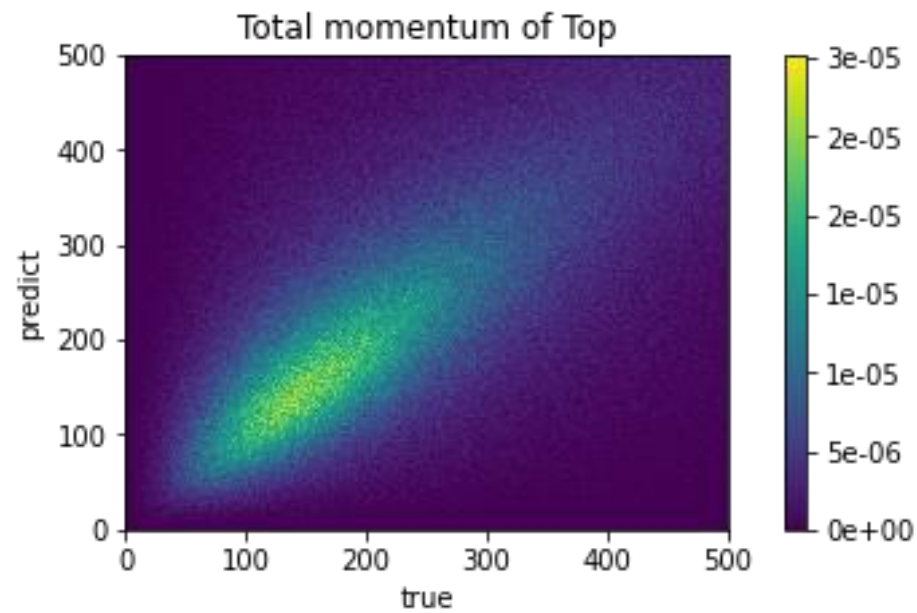
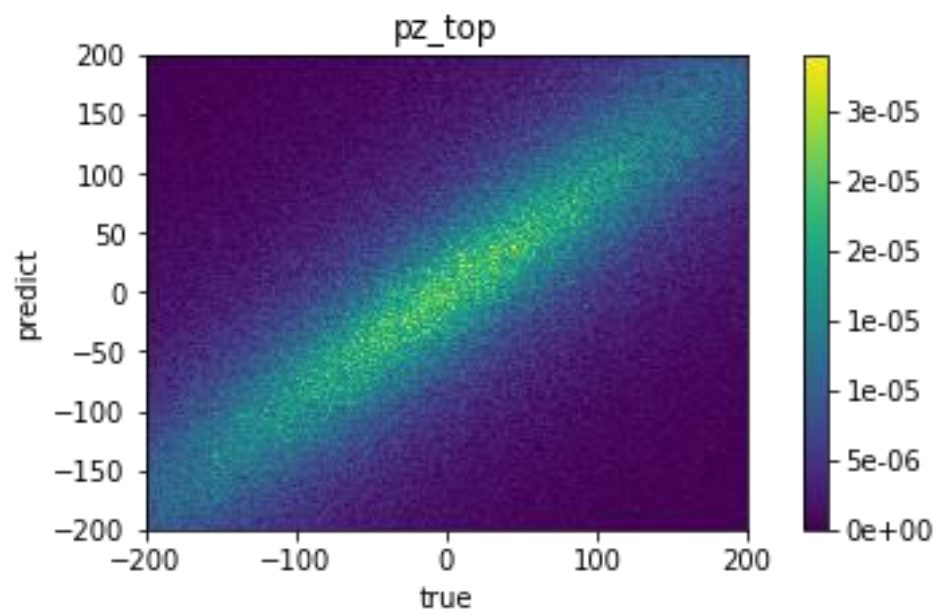
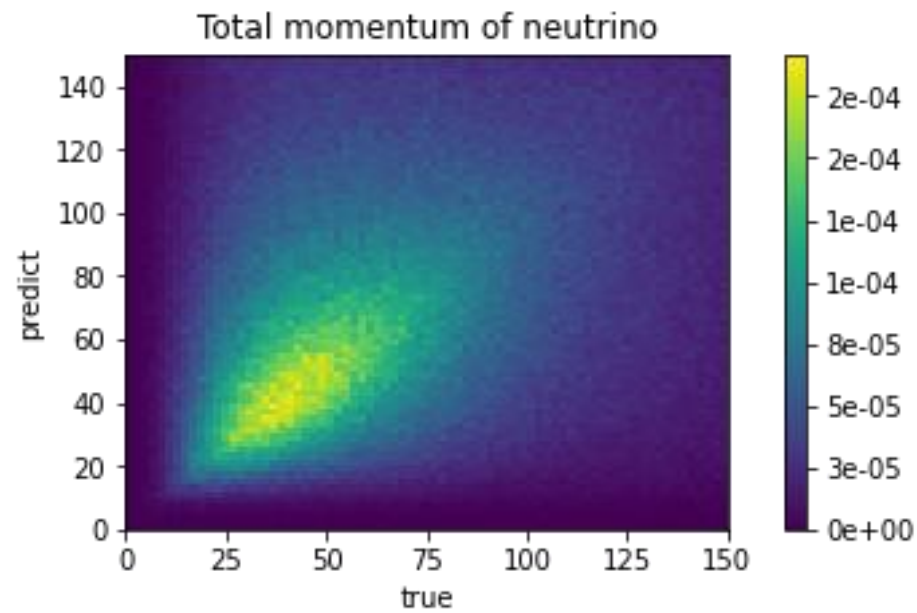
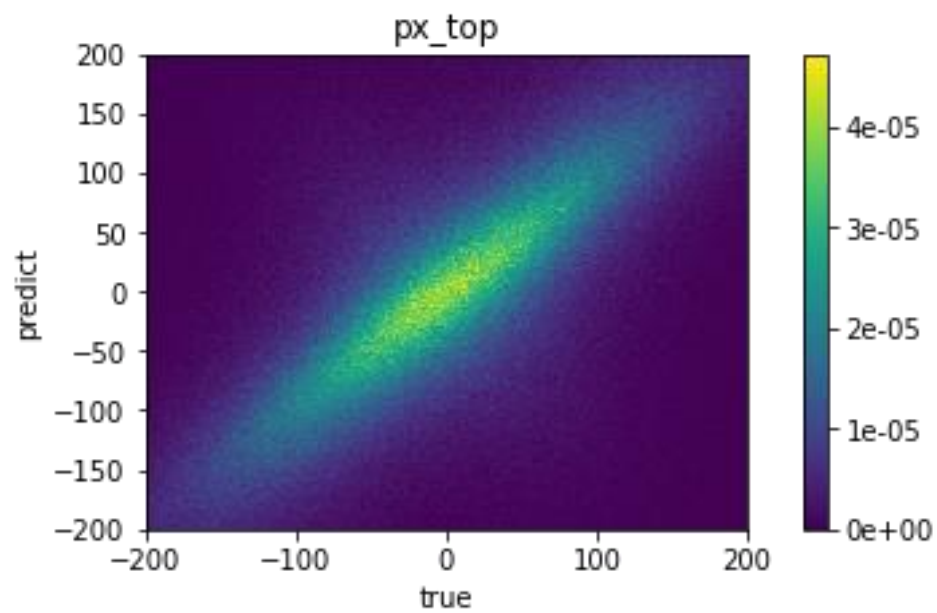
1812.09722 Erdmann, Geiser, Rath, Rieger



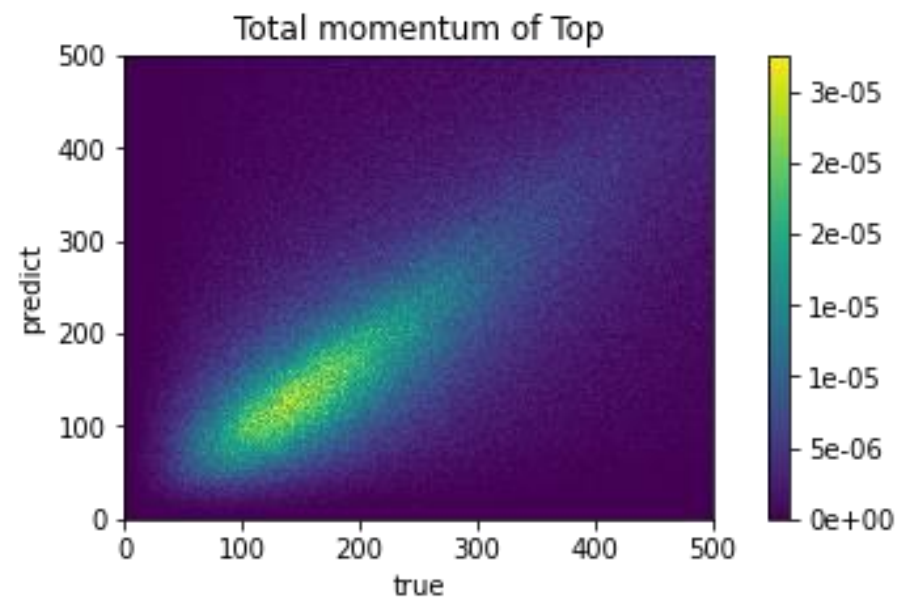
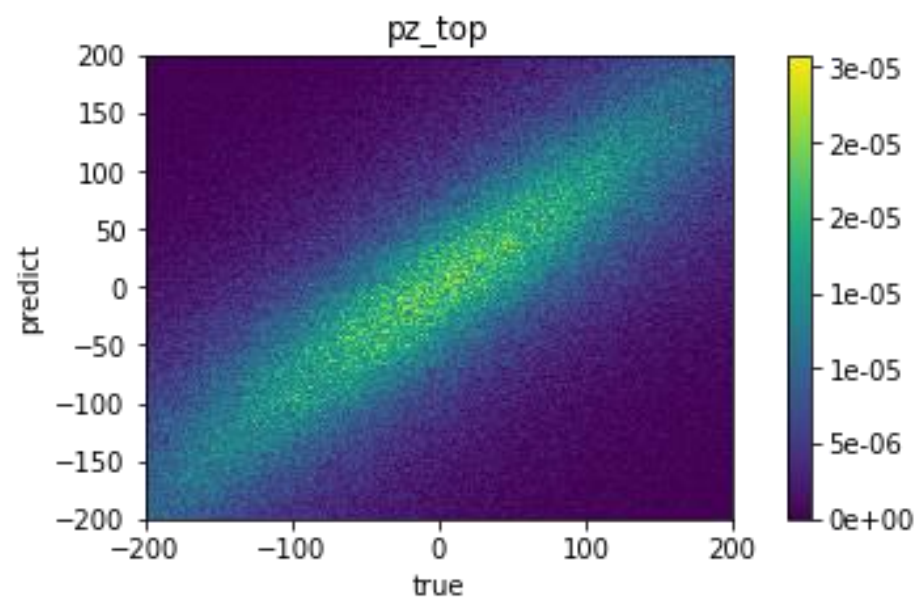
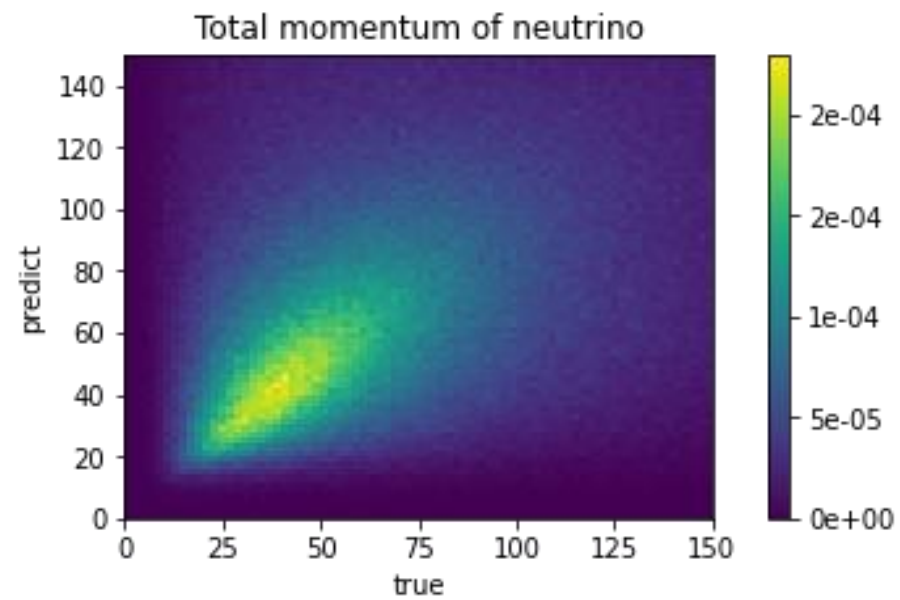
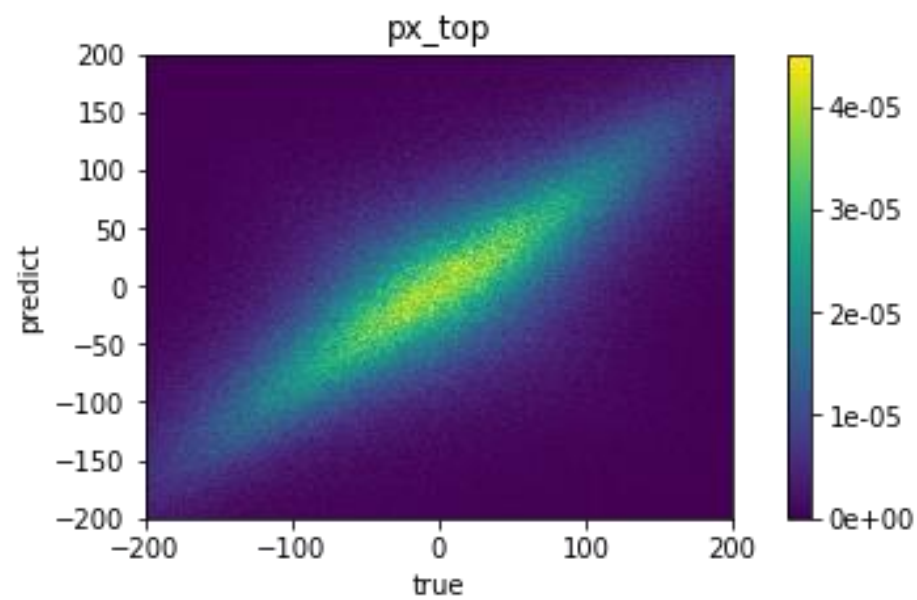
Analytic



M2cW



Neural Network



Chi squared

