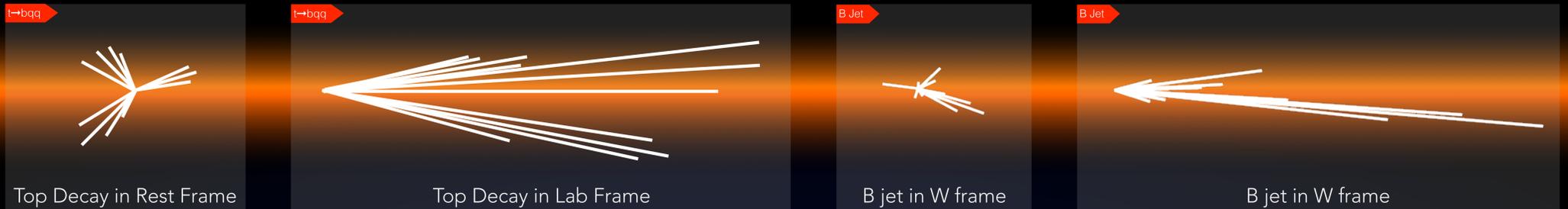


Identification of High-Momentum Top Quarks, Higgs Bosons, and W & Z Bosons using Boosted Event Shapes



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Motivation

LHC collaborations ATLAS and CMS have developed powerful new analysis techniques to identify high-momentum (“highly boosted”) hadronically decaying top quarks, Higgs bosons, and W & Z bosons, while rejecting background from multi-jet final states produced in QCD processes.

We present a new approach in identifying boosted massive particles. Our goal is a generic algorithm which can be used to determine the origin of individual “fat” jets with large transverse energy (E_T) produced in p-p collisions at the LHC.

Ideally this will optimize tagging t quarks, H bosons, and W & Z bosons, while discriminating against jets from light quarks and gluons.

Boosted Event Shapes

To generate Boosted Event Shapes (BES):

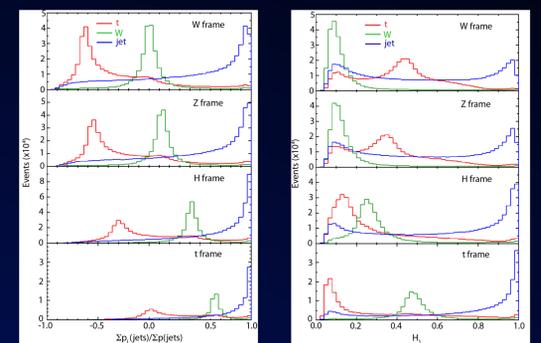
- Start with fat ($R=0.8$) jets
- Lorentz-boost jet constituents into t/H/Z/W frames

Calculate:

- Fox-Wolfram moments
- sphericity tensor eigenvalues
- thrust, aplanarity
- k_T jets in boosted frame
- In correct frame expect two or three jets, balanced momentum
- For each jet use 50 variables:
 - E_T , η
 - $H_1 - H_4$, $I_1 - I_3$, T , S_{pL} , S_p for each frame
 - $E_{T1} - E_{T4}$, m_{12} , m_{13} , m_{23} , m_{1-4} for W frame

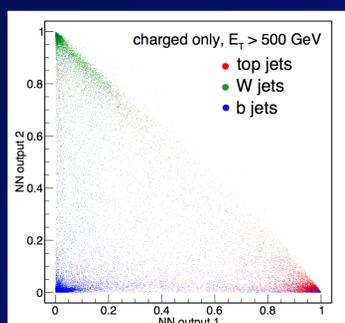
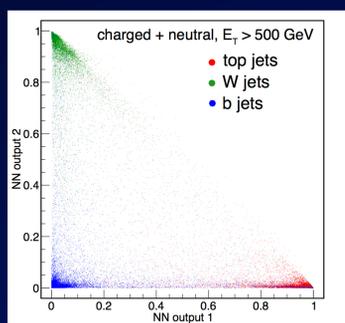
Simulations

- We use PYTHIA 3 TeV $Z' \rightarrow t\bar{t}$, HZ, W^+W^- , $b\bar{b}$
- Just used generator-level final state particles
- Impose 90% efficiency
- 2 GeV p_T threshold



Distributions of Fox-Wolfram moment H_1 (left) and ratio of longitudinal to total momentum of anti- k_T jets (right)

Neural Network BES Discriminator



ANN outputs of jets originating from all hadronic top quark decay, W decay, and b quarks for nets trained with charged & neutral (top) and only charged constituents (bottom)

Due to the large number of variables, we deploy artificial neural networks (ANNs) for the analysis. For tagging 3 jet types t, W, and b:

- Create training and test samples with 3.0 TeV Z' events
 - 50 variables per event
- Feed-forward net trained with vanilla back-propagation
 - learning strength 0.001; momentum 0.5
 - 3 outputs: b, W, and t jets
 - Stability achieved well before 1000 epoch training cycle
 - Varying layers & nodes minimally affected performance
- Tested BES algorithm with charged tracks only
 - more confusion between neutral particles at high momenta

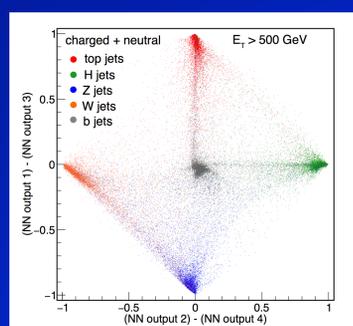


Also investigated was whether the new algorithm could tag all-hadronic top quark decays, H Boson $\rightarrow b\bar{b}$, $Z \rightarrow q\bar{q}$, and $W \rightarrow q\bar{q}$ when decaying particles are highly boosted.

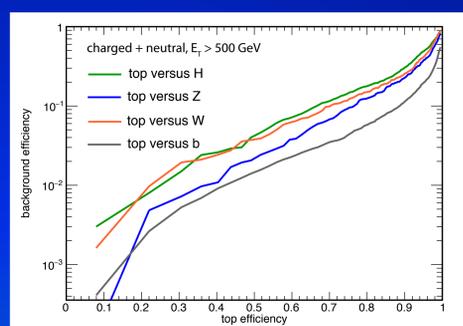
For tagging 5 jet types:

- Use PYTHIA again
- Pass through PGS
- 4 outputs: top, H, Z, W

We represent the results of this 5 jet problem as a 4 dimensional plot projected onto 2 dimensions.



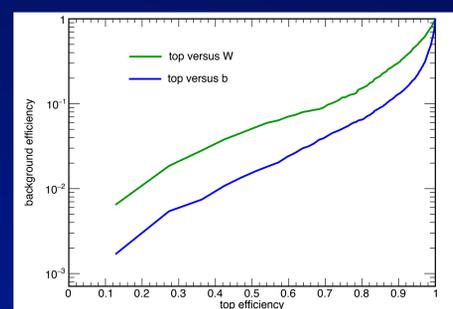
Distributions of ANN outputs for a 4-output net, tagging jets from all-hadronic top quark decay, H $\rightarrow b\bar{b}$ decay, Z $\rightarrow q\bar{q}$ decay, W $\rightarrow q\bar{q}$ decay, and b quarks



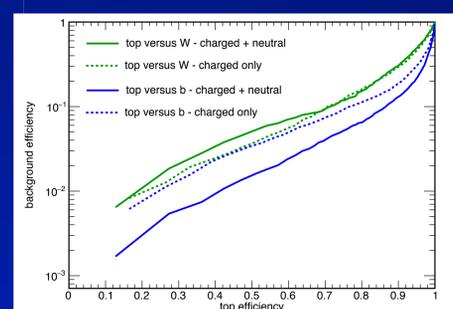
Efficiency of tagging jets from top decay (horizontal axis) versus efficiency for H, Z, W, and b jets

Generated:	t	W	b
Classified as t:	84.2%	3.9%	10.2%
Classified as W:	5.6%	86.8%	5.9%
Classified as b:	10.4%	9.3%	83.9%

Probabilities for classification of t, W, and b jets, while training on charged and neutral constituents.



Initial result ROC curves show background rejection versus top efficiency



Efficiency of tagging jets from top decay (horizontal axis) versus efficiency for W jets and b jets

[Event Shape Variables]

$$H_l = \sum_{i,j} \frac{|\mathbf{p}_i| |\mathbf{p}_j|}{E_{vis}^2} P_l(\cos\theta_{ij})$$

[Fox Wolfram Moments]

$$T = \max_{|\mathbf{n}|=1} \frac{\sum_i |\mathbf{n} \cdot \mathbf{p}_i|}{\sum_i |\mathbf{p}_i|}$$

[Thrust]

$$S^{\alpha\beta} = \frac{\sum_i p_i^\alpha p_i^\beta}{|\mathbf{p}_i|^2}$$

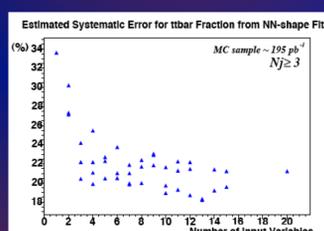
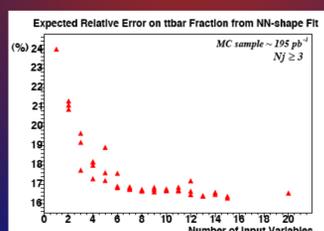
[Sphericity]

$$S = \frac{3}{2}(\lambda_2 + \lambda_3)$$

$$A = \frac{3}{2}\lambda_3$$

[Sphericity]

Worried about Systematics?



R. Marginean, Ph.D. thesis, Ohio State, CDF 2006

Summary

We have a new algorithm to identify very high momentum hadronically decaying top quarks, Higgs bosons, and vector bosons, by boosting into rest frames of hypothetical particle masses and then calculating a set of event shape quantities.

The algorithm is capable of separating jets which physically overlap in laboratory momentum space. In the idealized scenario, we see excellent separation of hadronic jets, largely maintained by using only charged jet constituents. Variables in this algorithm can be readily combined with other boosted particle tagging techniques.