Analyses

N-Subjettiness

HEPTopTagger

HOTVR

Resonances

Neural nets

## Top Tagging: Modern and Really Modern Methods

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Bangalore 1/2017

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## Fat jets

#### Boosted particles at the LHC [Steve's talk]

- 1994 boosted  $W \rightarrow 2$  jets from heavy Higgs [Seymour]
- 1994 boosted  $t \rightarrow 3$  jets [Seymour]



- 2002 boosted  $W \rightarrow$  2 jets from strongly interacting WW [YSplitter: Butterworth, Cox, Forshaw]
- 2006 boosted  $t \rightarrow 3$  jets from resonances [Agashe, Belyaev, Krupovnickas, Perez, Virzi]
- 2008 boosted  $H 
  ightarrow b ar{b}$  [BDRS Higgs tagger: Butterworth, Davison, Rubin, Salam]
- 2008 boosted  $t \rightarrow 3$  jets from resonances [JH/CMS tagger: Kaplan, Rehermann, Schwartz, Tweedie]
- 2009 boosted  $t \rightarrow 3$  jets in Higgs production [HEPTopTagger: TP, Salam, Spannowsky]
- 2009 boosted  $t \rightarrow 3$  jets from resonances [Template Tagger: Almeida, Lee, Perez, Sterman, Sung, Virzi]
- 2010 first meta analysis review [BOOST proceedings, Ed: Karagoz, Spannowsky, Vos]
- 2011 N-Subjettiness [Thaler, van Tilburg]

. . .

- 2011 Shower Deconstruction [Soper, Spannowsky]
- 2014 first machine learning study [Cogan, Kagan, Strass, Schwartzman]
- 2017 machine learning vs tagger t 
  ightarrow 3 jets [Kasieczka, Plehn, Russell, Schell]

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## Jet algorithms

### Definition of jets

- jet-parton duality  $\Leftrightarrow$  what are partons in detector?
- need algorithm to reconstruct what was one parton [IR save recombination algos]
- crucial for any LHC analysis [ILC analyses without jets?]
- extension to b and t perturbative QCD problem

#### Different measures [FASTJET: Cacciari, Salam, Soyez]

- define jet-jet and jet-beam distance [exclusive with resolution ycut]

$$k_{T} \qquad y_{ij} = \frac{\Delta R_{ij}}{R} \min \left( p_{T,i}, p_{T,j} \right) \qquad y_{iB} = p_{T,i}$$

$$C/A \qquad y_{ij} = \frac{\Delta R_{ij}}{R} \qquad y_{iB} = 1$$

$$anti-k_{T} \qquad y_{ij} = \frac{\Delta R_{ij}}{R} \min \left( p_{T,i}^{-1}, p_{T,j}^{-1} \right) \qquad y_{iB} = p_{T,i}^{-1}.$$

- (1) find minimum  $y^{\min} = \min_{ij}(y_{ij}, y_{iB})$ (2a) if  $y^{\min} = y_{ij}$  merge subjets *i* and *j*, back to (1) (2b) if  $y^{\min} = y_{iB}$  remove *i* from subjets, go to (1)
- theoretical and experimental trade-off decisions
- fat jets: use clustering history

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## Analysis: $Z' o t\bar{t}$

### LHC target $m_{Z'}\gtrsim$ 1.5 TeV

- purely leptonic decays rate limited
- semi-leptonic approximate reconstruction of neutrino 4-vector: massless neutrino 2D missing energy vector
  - top or W mass constraints
- purely hadronic decays deemed not useful

#### Many taggers [Hopkins/CMS tagger, HEPTopTagger, template tagger, shower deconstruction]

- hadronic top identification and reconstruction
- jet separation challenging for heavy Z'
- combination of calo and tracker great
- usually combined with b-tag

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### Validation and systematics

- tagging easier for higher boost,  $p_{T,t} > 600 \text{ GeV}$
- Standard Model events at lower  $p_{T,t}$  < 400 GeV
- $\Rightarrow p_T$  range main challenge



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## Analysis: $t\bar{t}H, H \rightarrow b\bar{b}$

### Challenge of $t\bar{t}H, H \rightarrow b\bar{b}$ [CMS-TDR]

- trigger:  $t \to bW^+ \to b\ell^+\nu$ reconstruction and rate:  $\overline{t} \to \overline{b}W^- \to \overline{b}jj$
- continuum background ttbb, ttjj [weighted by b-tag]
- 1- combinatorics:  $m_H$  in  $pp \rightarrow 4b_{tag}$  2j  $\ell \nu$ 2- kinematics: peak-on-peak 3- systematics:  $S/B \sim \mathcal{O}(0.1)$

Tagger [TP, Salam, Spannowsky; MadMax: TP, Schichtel, Wiegand]

- boost solving combinatorics [buckets]
- significance known to be in boosted regime
- side band in m<sub>bb</sub> possible
- fun: fat Higgs jet + fat top jet



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### Stop pairs including reconstruction [TP, Spannowsky, Takeuchi, Zerwas]

- stop crucial for hierarchy problem comparison to other top partners
- hadronic:  $\tilde{t}\tilde{t}^* \rightarrow t\tilde{\chi}^0_1 \ \bar{t}\tilde{\chi}^0_1$

Analysis: top partners

- stop mass from m<sub>T2</sub> endpoint [like sleptons or sbottoms]

$$m_{T2}(\hat{m}_{\chi}) = \min_{\substack{\phi_T = q_1 + q_2}} \left[ \max_j m_{T,j}(q_j; \hat{m}_{\chi}) \right] \stackrel{!}{<} m_{\tilde{t}}$$





#### Analyses

#### N-Subjettiness

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## N-Subjettiness

### N-Jettiness to count subjets inside fat jet [Thaler, van Tilburg; Jesse's talk?]

- how many subjets do the calo entries correspond to?
- event shape using N subjet directions  $\hat{n}_{j}$  [ $\beta > 0$ ]

$$\tau_{N} = \frac{1}{\sum_{\alpha \in jet} p_{T,\alpha} R_{0}^{\beta}} \sum_{\alpha \in jet} p_{T,\alpha} \min_{k=1,\ldots,N} (\Delta R_{k,\alpha})^{\beta}$$

- choice of reference axes
  - 1- from subjet algorithm
  - 2- from minimization of  $\tau_N$

# $-\tau_N \rightarrow 1$ means many calo entries away from *N* axes $\tau_N \rightarrow 0$ means perfect matching

- systematics cancelled in ratios  $\tau_{N+1}/\tau_N$  with dip for N+1 subjets



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### Tagger

- simple selection

$$m_{\rm fat \, jet} = 160...240 \ {
m GeV} \qquad { au_2 \over au_2} < 0.6$$

- multi-variate in  $\pmb{N}$  and  $\beta$  with some improvement
- $\Rightarrow$  easily added to any other tagger

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#### HEPTopTagger

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## HEPTopTagger

### Mass drop algorithm [TP, Salam, Spannowsky, Takeuchi]

- 1- C/A fat jet, R = 1.5 and  $p_T > 200 \text{ GeV}$  [FastJet limitation]
- 2– mass drop, cutoff  $m_{sub} > 30 \text{ GeV}$
- 3- filtering leading to hard substructure triple
- 4– top mass window  $m_{123} = [150, 200]$  GeV
- 5– A-shaped mass plane cuts as function of  $m_W/m_t$
- 6– consistency condition  $p_T^{(tag)} > 200 \text{ GeV}$







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### Top reconstruction

- direction less critical
- energy requiring calibration





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### Limitations to watch

- signal efficiency
- background sculpting
- $-p_T$  range
- resonance reconstruction

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## A masterpiece in diplomacy

### ATLAS resonance search [CERN-PH-EP-2012-291, ATLAS-CONF-2013-084]

- resonances decaying to  $t\bar{t}$  [mass 1.0-1.5 TeV]
- $\mathit{Z'}$  or KK gluon only different in width,  $\Gamma_{\mathit{Z'}} < \Gamma_{\mathit{G}}$
- semi-leptonic searches done before
- $\Rightarrow$  mostly test of top taggers

### Subjet methods [Kasieczka, Schätzel, Anders (ATLAS Heidelberg)]

- starting with lots of jet calibration
- also add b-tag in/around fat jet
- fat jet and top masses in data [background region]



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- $\Rightarrow$  top tagger basics established

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#### Resonance search

- *m*<sub>tt</sub> from TemplateTagger and HEPTopTagger



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### Resonance search

- *m*<sub>tt</sub> from TemplateTagger and HEPTopTagger
- limit on Z'
- limit on KK gluon
- $\Rightarrow$  it all works!



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#### HEPTopTagger

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## Better HEPTopTagger

Multivariate BDT setup [TP, Spannowsky, Takeuchi; Anders, Bernaciak, Kasieczka, TP, Schell]

- increase size R = 1.8
- change in cut order  $\mathbf{3}\leftrightarrow \mathbf{4}$
- alternative triplet selection

maximal  $dj_{sum} = \sum p_{T,i} p_{T,j} (\Delta R_{ij})^4$  triplet of hardest subjets

 $- \{m_{13}, m_{23}, m_{123}\}$ 





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#### Moderate boost [Anders, Bernaciak, Kasieczka, TP, Schell]

- focus on  $p_T = [150, 200]$  GeV
- target type-2 tags [two hardest subjets matched]
- correlations via Fox-Wolfram moments

$$H_{\ell}^{U} = \sum_{i,j} \frac{1}{N^2} P_{\ell}(\cos \Omega_{ij})$$

 $- \{m_{13}, m_{23}, m_{123}, FWMs\}$ 



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## HEPTopTagger2

#### OptimalR [Kasieczka, TP, Salam, Schell, Strebler]

- optimal fat jet size Rmin [large to include decay jets, small to avoid combinatorics]
- reduce R until we leave jet mass plateau

 $|m_{123} - m_{123}^{(R_{\text{max}})}| < 0.2 m_{123}^{(R_{\text{max}})} \Rightarrow R_{\text{min}}$ 

- estimate  $R_{\min}^{(calc)}$  from kinematics

$$- \{m_{123}^{(R_{\min})}, f_{W}^{(R_{\min})}, R_{\min} - R_{\min}^{(\text{calc})}\} [f_{W} = \min[m_{ij}/m_{123} - m_{W}/m_{t}]]$$

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### N-Subjettiness [Thaler, van Tilburg]

- also include rejected events  $[R_{filt} = 0.2, N_{filt} = 5]$
- $\{\boldsymbol{m}_{123}^{(\boldsymbol{R}_{\min})}, \boldsymbol{f}_{\boldsymbol{W}}^{(\boldsymbol{R}_{\min})}, \boldsymbol{R}_{\min} \boldsymbol{R}_{\min}^{(\text{calc})}, \tau_{j}, \tau_{j}^{(\text{filt})}\}$

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#### Qjets [Ellis, Hornig, Roy, Krohn, Schwartz]

- more than one clustering history, weighted by

$$\begin{split} \omega_{ij} = \exp\left[-\alpha \frac{y_{ij} - y_{ij}^{\min}}{y_{ij}^{\min}}\right] \\ \text{then using distributions like } \langle m^2 \rangle - \langle m \rangle^2 \\ - \left\{m_{123}^{(\mathcal{R}_{\min})}, f_{W}^{(\mathcal{R}_{\min})}, \mathcal{R}_{\min} - \mathcal{R}_{\min}^{(\text{calc})}, \tau_j, \tau_j^{(\text{filt})}, \left\{m_{123}^{\text{Qjets}}\right\}\right\} \end{split}$$

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### we have been a second

Non-improvements

- Tagging without fat jets
   [Lapsien, Kogler, Haller; FastJetContrib]
- 1- variable-R: jet size depending on  $p_T$  [Krohn, Thaler, Wang]
- 2– mass jump: jet mass condition in C-A jet algorithm [stoil] specifically  $m_{ij} \lesssim 30 \text{ GeV}$  and  $m_{ij} \gtrsim 1.4 \max(m_{ij})$ 
  - possible advantage: not relying on fixed fat jet
  - comparison with HEPTopTagger, CMSTagger in working points pseudo-ROC curve by adding \(\tau\_{2,3}\) [why???]



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- $\Rightarrow$  tagger performance stable and plateau'd
- $\Rightarrow$  find other ways to improve?

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#### Resonances

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### Resonance search

### Final state radiation

- Z' search without ISR/FSR



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#### Resonances

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- Z' search without ISR/FSR
- FSR major problem



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### Resonance search

### Final state radiation

- Z' search without ISR/FSR
- FSR major problem
- using reconstructed 4-momentum of tagged top
- tail in  $m_{tt}^{(rec)}$  from tagged momentum and FSR
- $\Rightarrow$  add 4-momentum of fat jet: { $m_{tt}, p_{T,t}, m_{jj}^{\text{(filt)}}, p_{T,j}^{\text{(filt)}}$ }



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### Full blast HEPTopTagger2

- $-~Z' 
  ightarrow t ar{t}$  signal
- high efficiencies remaining a challenge
- $\Rightarrow$  ball in experimenalists' field



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## Neural networks

### Recent developments in image regonition [convolutional networks]

- wavelet transformation [Rentala, Shepherd, Tait; Monk]
- W-tagging with image recognition [Cogan etal, Oliveira etal, Baldi etal]
- top-tagging attempt [Almeida, Backovic, Cliche, Lee, Perelstein]
- QCD and shower study [Barnard etal]
- quark-gluon discrimination [Komiske etal]
- ⇒ does image recognition beat taggers? [Kasieczka, TP, Russell, Schell]

Start with anti- $k_T$  – 1.5 fat jet [ $p_T$  = 350...450 GeV,  $\Delta \eta$  = 0.1 vs  $\Delta \phi$  = 5°]

- find maxima identify characteristic points
- shift move image to center the global maximum
- rotation rotate the second maximum to 12 o'clock
- flip ensure third maximum is in the right half-plane
- crop crop the image to  $40 \times 40$  pixels
- decide on E vs  $E_T$  for rapidities  $\eta\gtrsim$  2
- $\Rightarrow$  all not strictly necessary, but obviously CPU-saving



- Analyses
- HEPTopTagge
- HOTVR
- Resonances
- Neural nets

### Getting binned calorimeter images ready $[n_{\eta} \times n_{\phi} \text{ energy bins}]$

- Convolutional layers to transform into optimal basis with kernel matrix W
- Activation to weigh the feature maps  $\vec{x}$ , for example step function
- Pooling to reduce the size of picture

Neural networks

- ConvNet output 6 kernels W, 4 layers/iterations



### Analyses N-Subjettin

- HOTVB
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### Neural networks

-

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- DNN output to link to data  
$$y_i = \max\left(0, \sum_{j=1}^{n^2} W_{ij}x_j + b_i\right) \quad \text{for} \quad i \in [1, n_{d-node}]$$

- loss function to train 
$$\{W, b\}$$
 by minimization  

$$L(\theta) = \frac{1}{N} \sum_{i=0}^{N} (y(W, b; x_i) - y_i)^2$$

- minimization with learning rate, momentum

$$\theta_{n+1} = \theta_n - \eta \nabla L(\theta_n) + \alpha(\theta_n - \theta_{n-1})$$



Analyses

- N-Subjettines
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### Compare ConvNet with QCD tagger — preliminary! [including N-subjettiness, no Qjets]

1- state of the art HEPTopTagger2

Performance

2- ConvNet trained and applied on same simulation environment

Analyses N-Subjettines HEPTopTagg HOTVR

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### Performance

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 $\Rightarrow$  QCD is safe, thinking and understanding hard to beat

Analyses N-Subjettine HEPTopTag HOTVR

Neural nets

### Outlook

### I am probably over time, so...

- ...QCD-based taggers are stable
- ...information beyond clustering history is needed
- ...some progress does not survive v2
- ...imagine recognition is cool, but QCD is cooler

### Thank you to ...

...former and current ATLAS-Heidelberg ...all groups working with and on taggers

Lectures on LHC Physics, Springer, arXiv:0910.4182 updated under www.thphys.uni-heidelberg.de/~plehn/