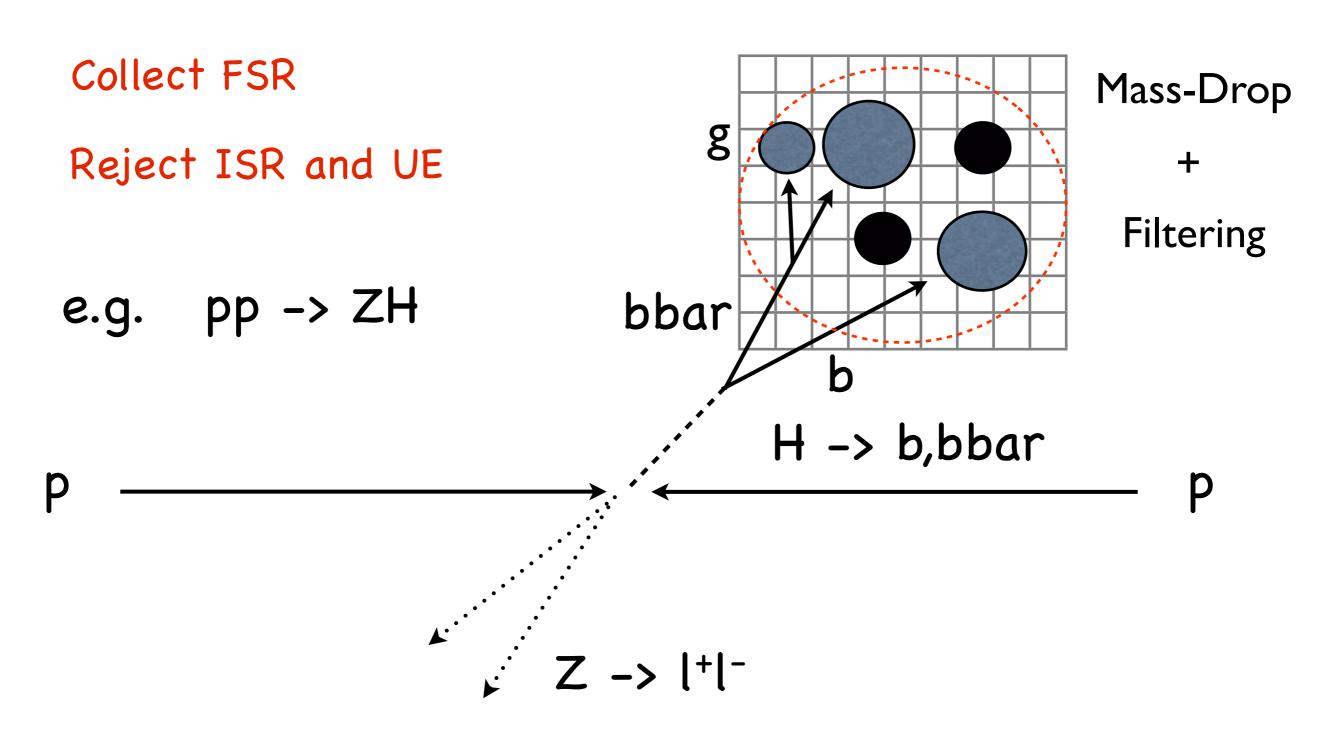
Boosted Higgs search

Michael Spannowsky

ITS, University of Oregon

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Basic idea of subjet analysis

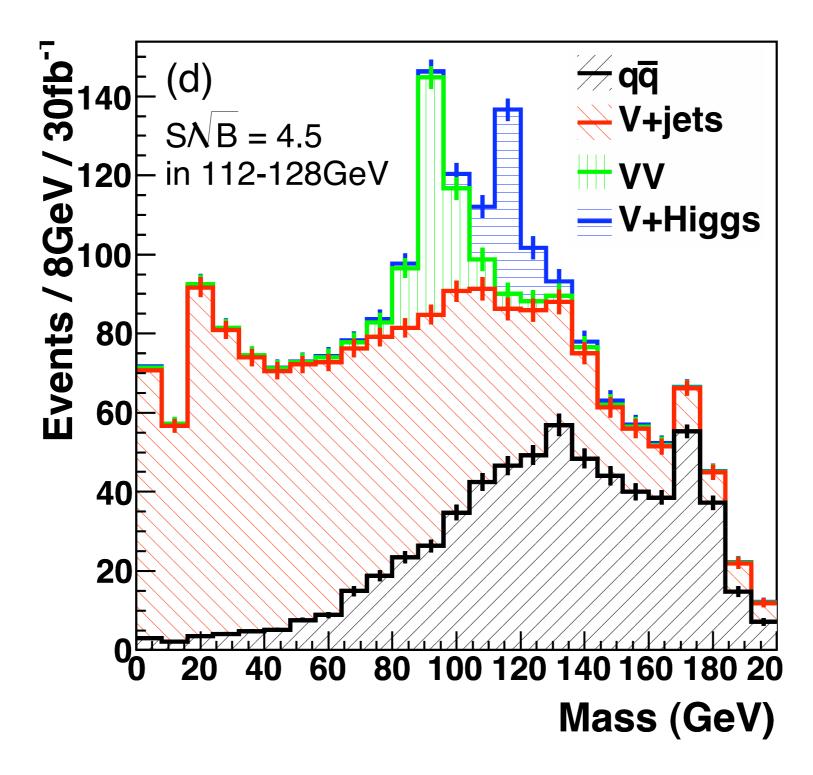


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Result for ZH and WH combined:



Confirmed by ATLAS analysis with only slightly smaller significance [ATL-PHYS-PUB-2009-088, G. Piacquadio]

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Further techniques for jet grooming

Pruning

[PRD 80 (2009), S.Ellis, Vermilion and Walsh]

• Run CA or kT algorithm on constituents of found jet (large cone: R). At each recombination $i, j \rightarrow p$ test if:

$$z = rac{\min(p_{T_i}, p_{T_j})}{p_{T_p}} < z_{ ext{cut}}$$
 and $\Delta R_{ij} > D_{ ext{cut}}$

- If both conditions are full-filled,
 veto on the recombination discard
 the lower pT daughter and
 continue.
- The resulting jet is the new (pruned) jet

Trimming

[JHEP 1002 (2010), Krohn, Thaler and Wang]

- Run a jet algorithm (kT) on a found jet (Anti-kT, large cone R), with Rsub smaller than R
- ullet Discard all subjets with $\,p_T = f_{
 m cut} \Lambda$
- Remaining subjets from the new (trimmed) jet

Pruning vs Trimming

David Krohn's conclusion at the Workshop in Seattle:

- * Trimming and Pruning are complementary techniques:
 - Trimming is for QCD jets, Pruning for boosted heavy objects

http://silicon.phys.washington.edu/JetsWorkshop/Krohn.pdf



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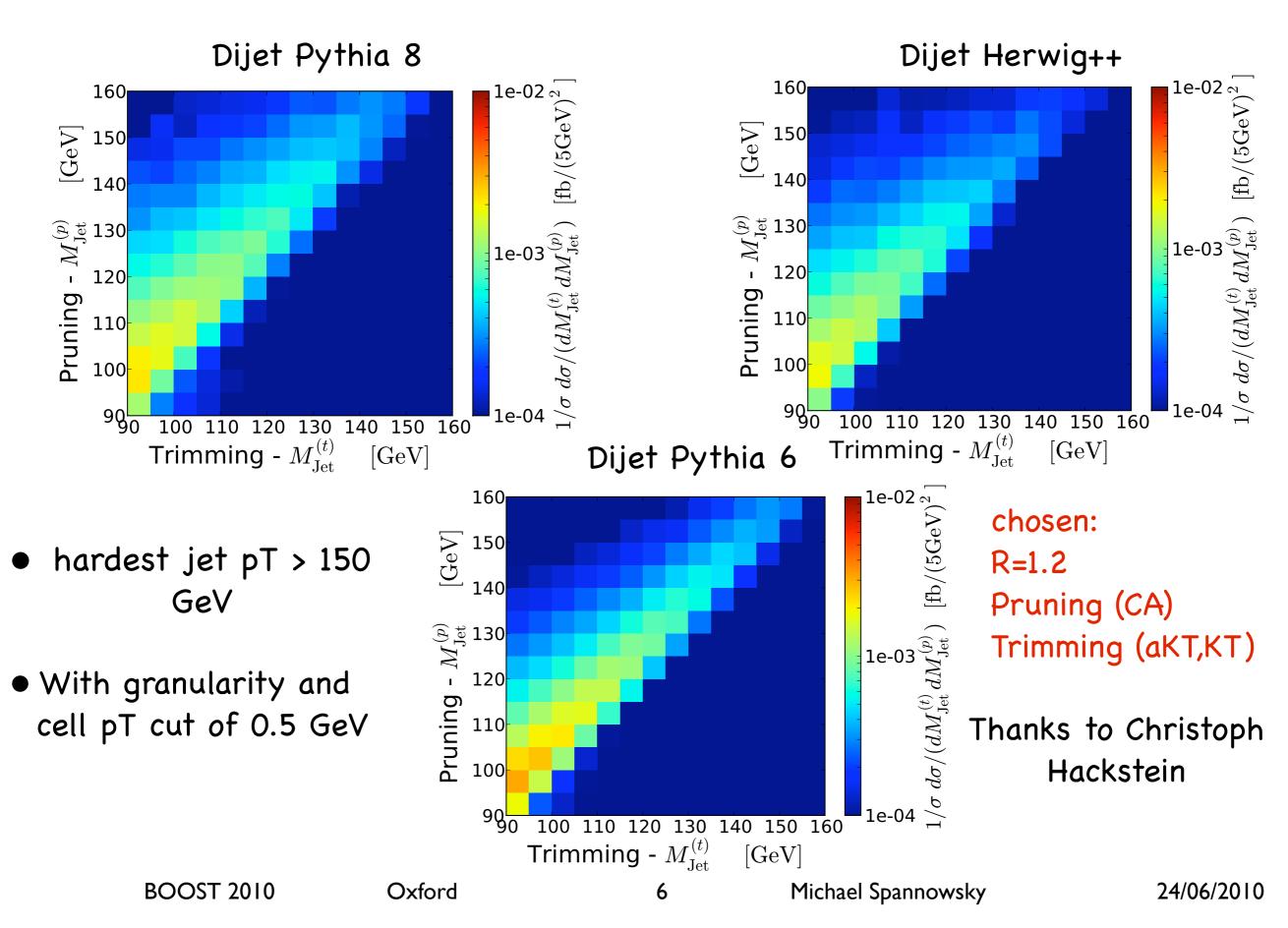


Both do their job but differently

Use it!

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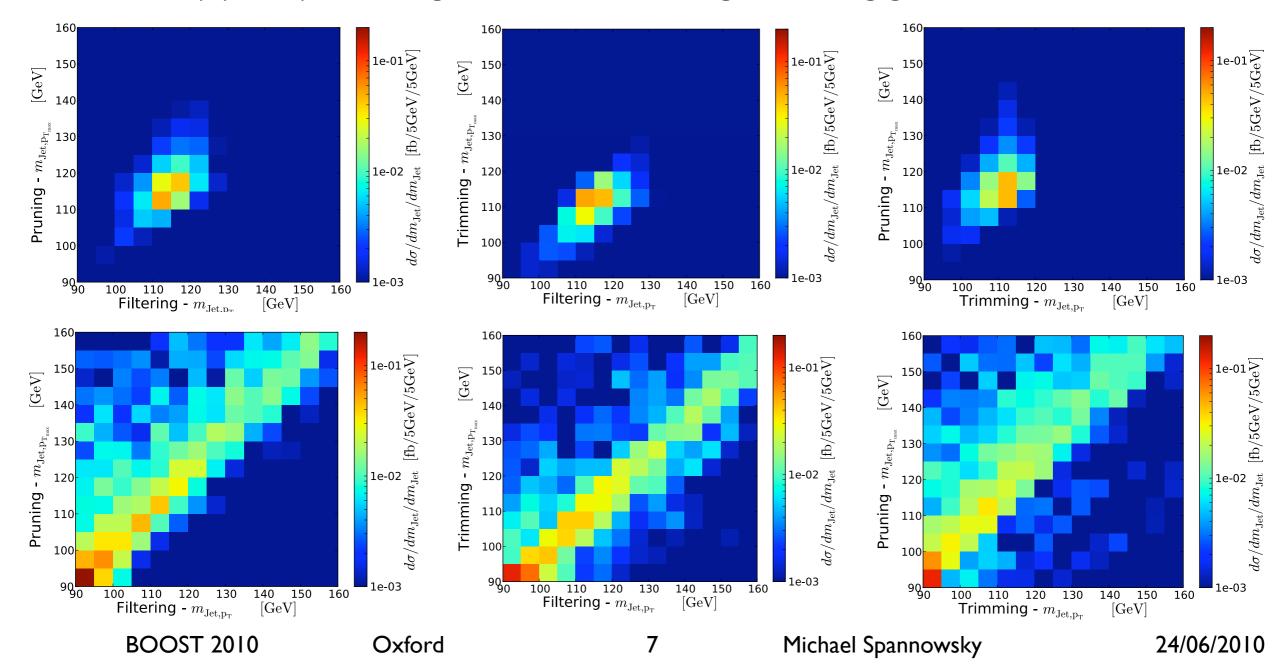
What happens if we run Pruning and Trimming simultaneously?



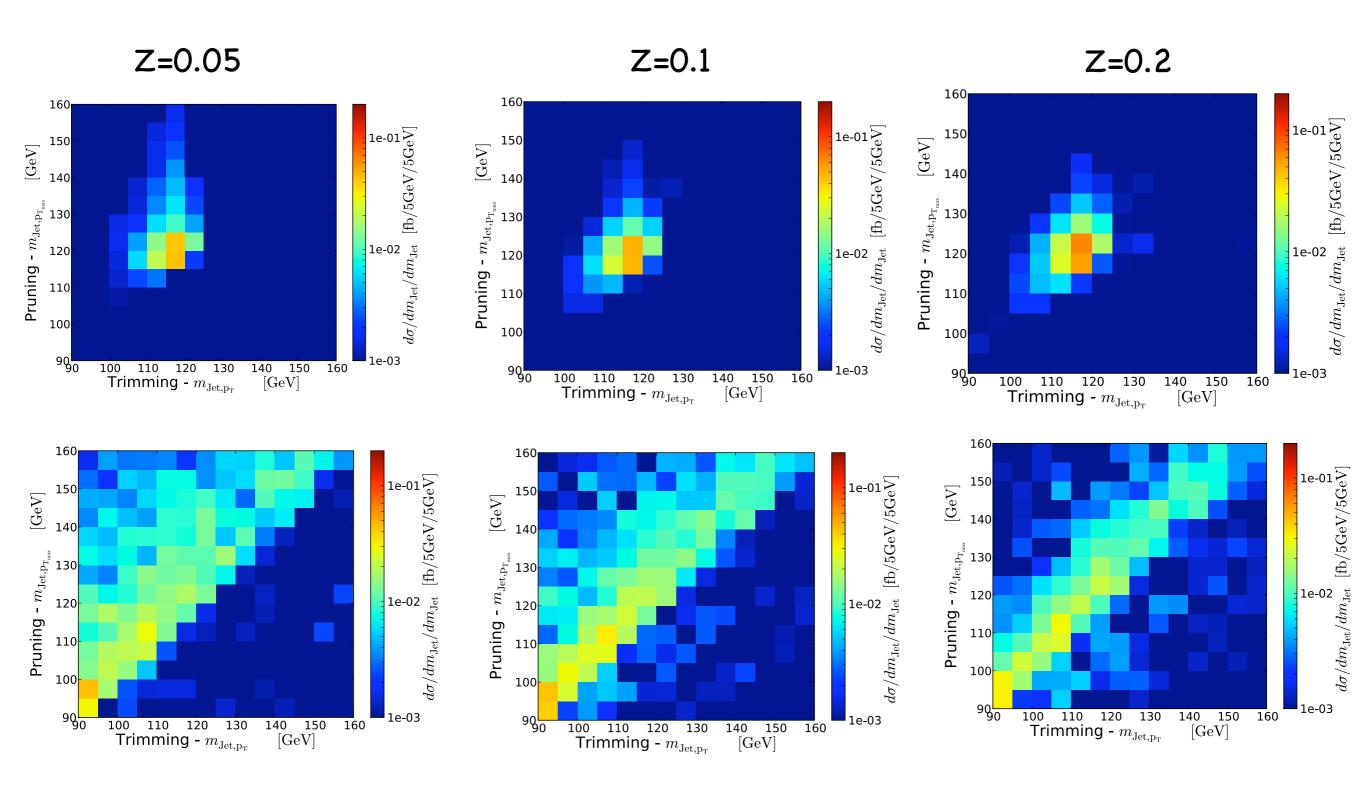
Combine Pruning, Trimming and Mass-Drop/Filtering

[Soper, MS, 1005.0417]

Run BDRS analysis for ZH production Apply pruning and trimming on Higgs candidate



'Optimize' procedure



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Exploitation of asymmetry

Cut based approach

Exp. Likelihood Ratio
$$\langle \mathcal{L}(\{n\})\rangle_{\mathrm{SB}} = \sum_J \left[(s_J + b_J) \log \left(1 + \frac{s_J}{b_J} \right) - s_J \right]$$

| | $M_{ m Jet}^{(f)} \in W_f$ | $M_{\text{Jet}}^{(f)} \in W_f$ $M_{\text{Jet}}^{(t)} \in W_t$ | $M_{\text{Jet}}^{(f)} \in W_f$ $M_{\text{Jet}}^{(p)} \in W_p$ | $M_{\text{Jet}}^{(p)} \in W_p$ $M_{\text{Jet}}^{(t)} \in W_t$ | $M_{\text{Jet}}^{(p)} \in W_p$ $M_{\text{Jet}}^{(t)} \in W_t$ |
|--|----------------------------|---|---|---|---|
| Signal cross section [fb] | 0.20 | 0.18 | 0.17 | 0.17 | 0.16 |
| Backgrnd cross section [fb] | 0.30 | 0.20 | 0.17 | 0.16 | 0.13 |
| s/b | 0.67 | 0.90 | 1.0 | 1.1 | 1.3 |
| s/\sqrt{b} $(\int dL = 30 \text{ fb}^{-1})$ | 2.0 | 2.2 | 2.3 | 2.3 | 2.4 |
| $\langle \mathcal{L}(n) \rangle_{\rm SB} \ (\int dL = 30 \text{ fb}^{-1})$ | 1.7 | 1.9 | 2.0 | 2.1 | 2.2 |

Z=0.1 Z=0.05

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Z=0.1 Z=0.05

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Stronger as a team

Filtering

Trimming

Pruning



Pruning

Trimming

Filtering

Generic resonance tagger should run comb. of procedures Maybe we can gain insight to improve on subjet procedures

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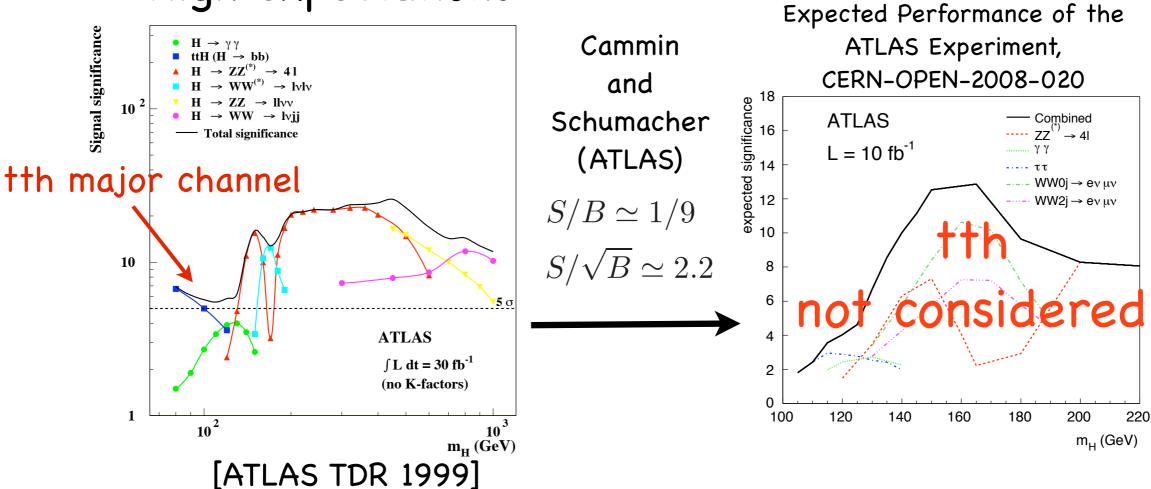
Brief summary of tth

[Plehn, Salam, MS, PRL 104 2010]

Motivation:

- sizable cross-section
- Higgs discovery contribution in low mass range
- access to t- and b-Yukawa couplings

High expectations:



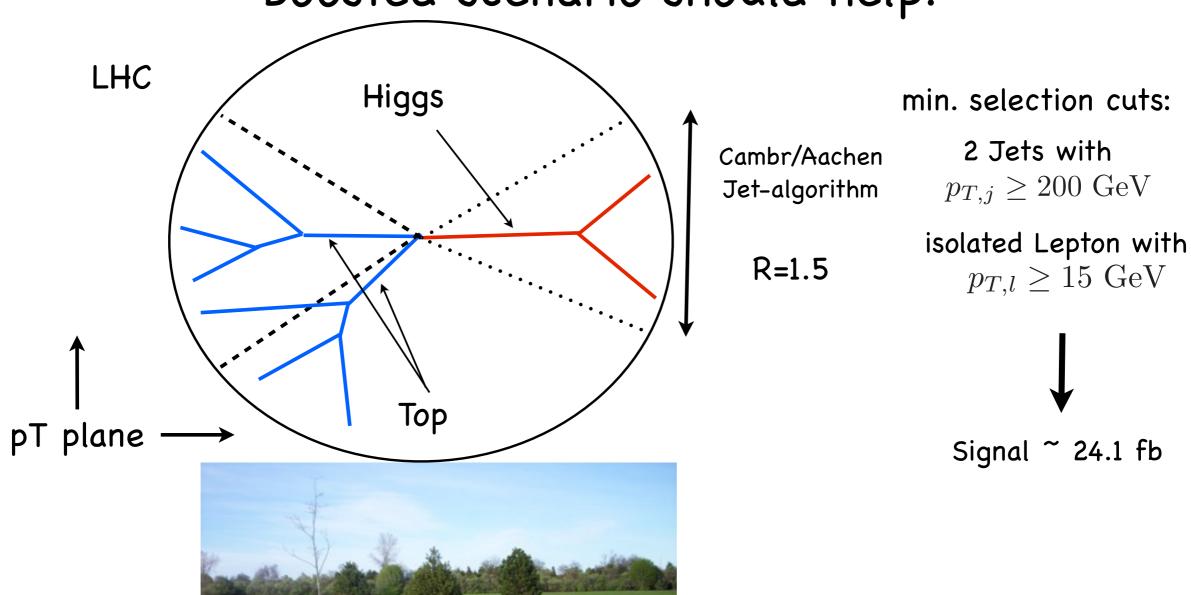
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Boosted scenario should help!



Only 2 or 3 b in one cone reduces combinatorics

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Nasty backgrounds.....

tth (Signal)

Beenakker et al.,
PRL 87 2001;
Reina et al.,
PRD 65 2002

→ K=1.57

ttbb

Bredenstein et al., PRL 103 2009; Belivacqua et al., JHEP 0909 2009

► K=2.3

tt+jets

Dittmaier et al, PRL 98 2007 Bevilacqua et al., PRL 104 2010

► K=1.0

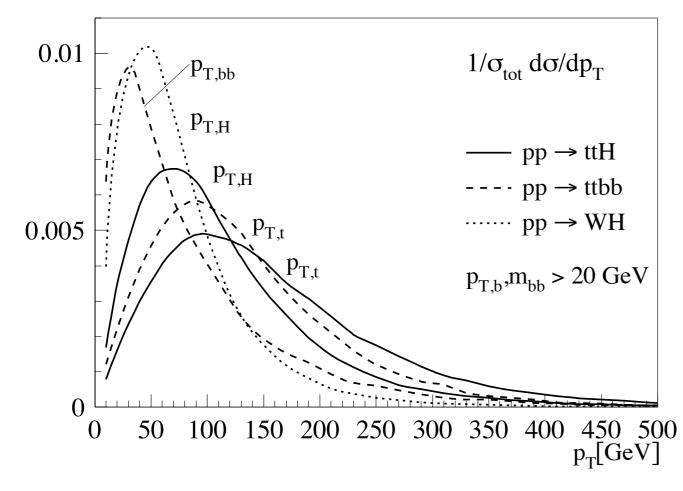
ttz

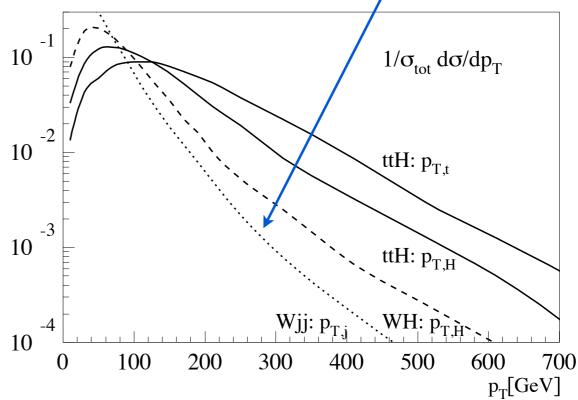
Lazopoulos et al., PLB 666 2008

→ k=1.53

w+jets

negligible after b-tags and taggers





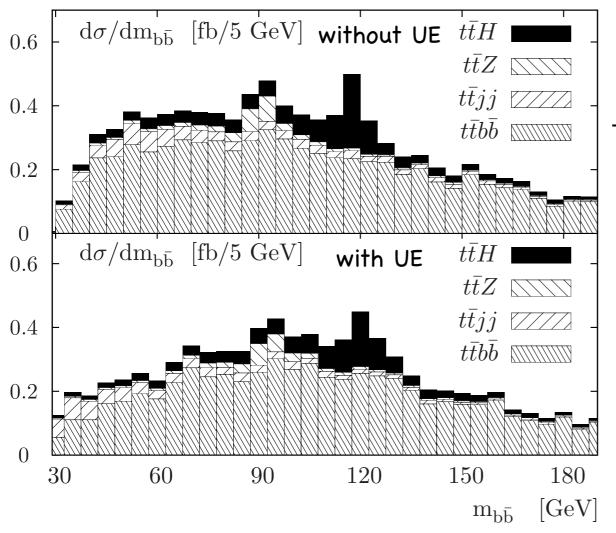
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Results



for 100 1/fb

with 2 b-tags

$$S[fb^{-1}]$$
 $B[fb^{-1}]$
 S/B
 S/\sqrt{B}
 $m_H = 115 \text{ GeV}$
 1.2
 3.8
 $1/3.2$
 6.2
 120 GeV
 1.0
 3.8
 $1/3.8$
 5.1
 130 GeV
 0.51
 3.3
 $1/6.5$
 2.8

- → tremendous improvement on S/B in tth
- → tth might contribute to Higgs discovery
- → tth might be a window to Higgs-top coupling

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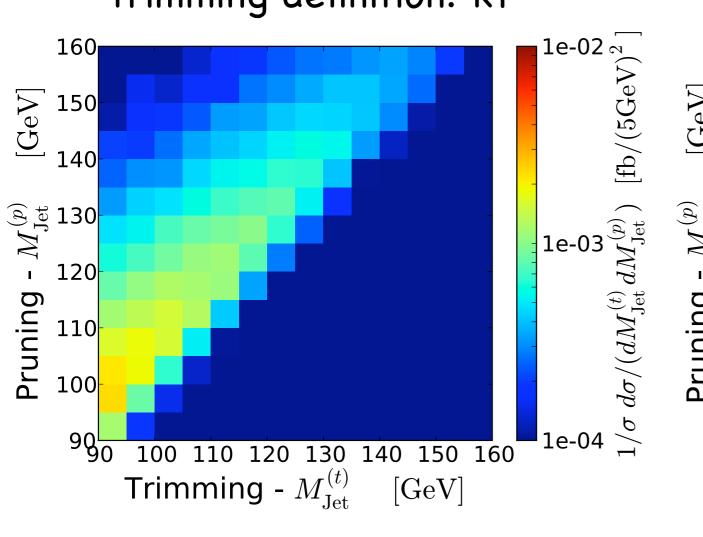
Backup

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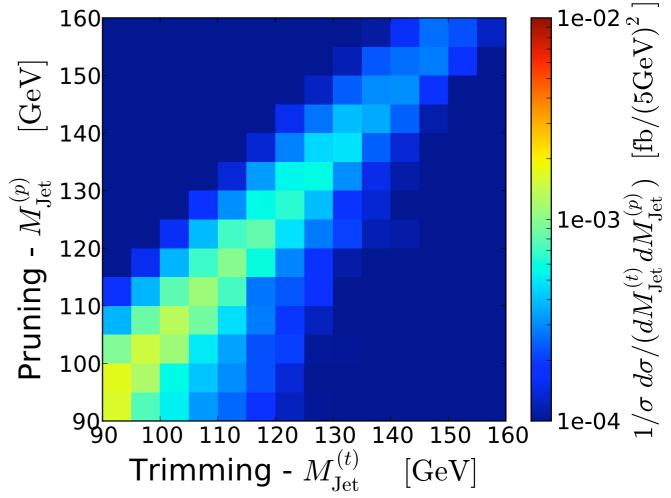
Pruning vs Trimming for same fat-jet definition

fat jet definition: CA pruning definition: CA

trimming definition: kT



fat jet definition: anti-kT pruning definition: CA trimming definition: kT



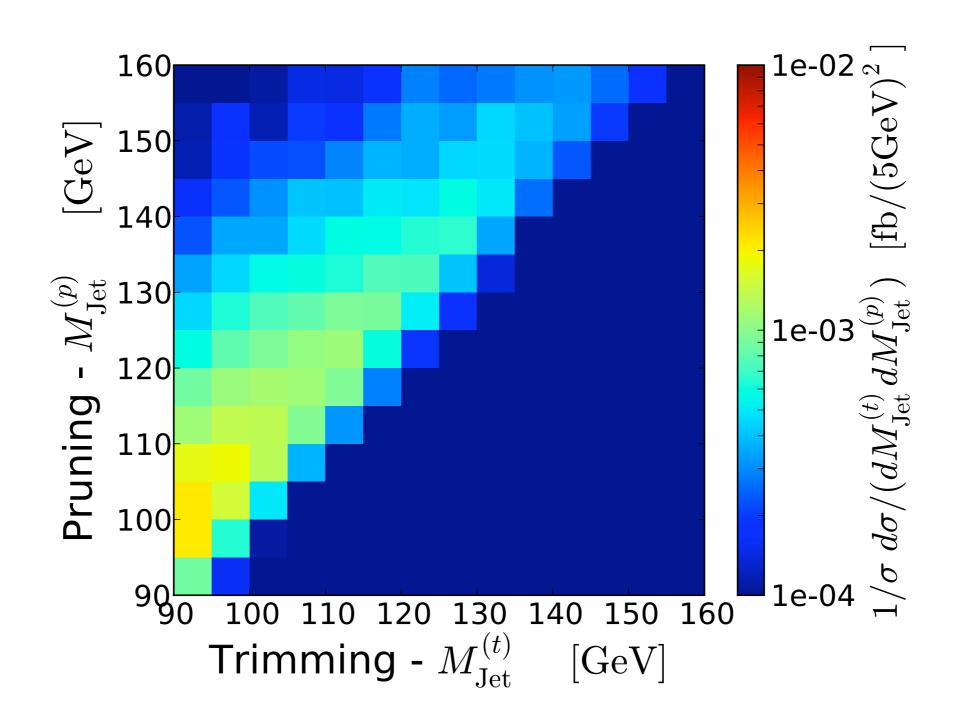
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Pruning and Trimming using CA



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Log Likelihood Ratio

Poisson distributions for background and signal + background:

$$P_{\rm B}(\{n\}) = \prod_{J} \frac{1}{n_J!} (b_J)^{n_J} e^{-b_J} \qquad P_{\rm SB}(\{n\}) = \prod_{J} \frac{1}{n_J!} (b_J + s_J)^{n_J} e^{-b_J - s_J}$$

Likelihood ratio

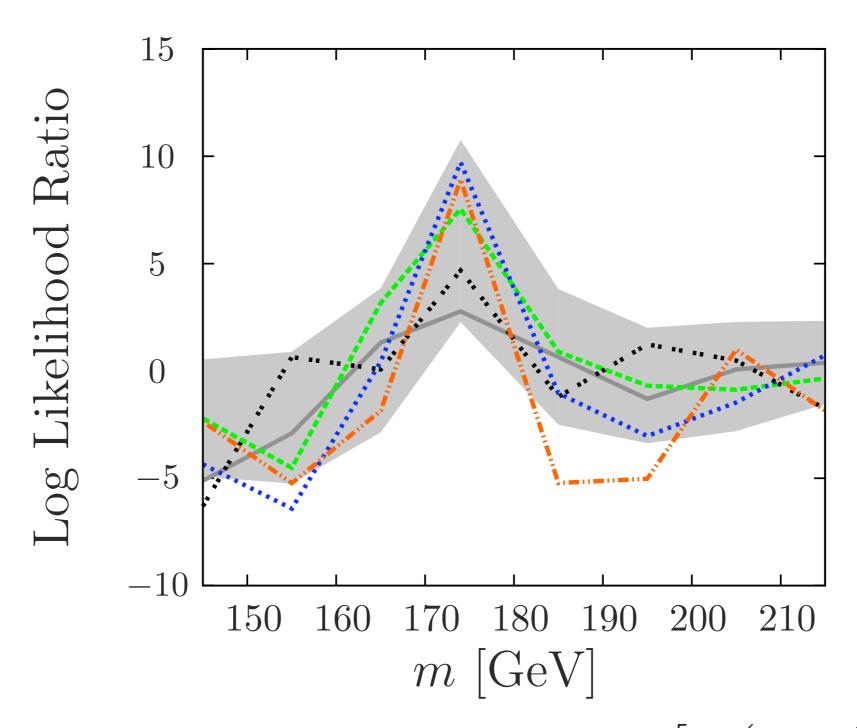
$$R(\{n\}) = \frac{P_{SB}(\{n\})}{P_{B}(\{n\})}$$

$$R(\{n\}) = \frac{P_{\text{SB}}(\{n\})}{P_{\text{R}}(\{n\})}$$
 $R(\{n\}) = \exp \mathcal{L}(\{n\})$

Log Likelihood ratio

$$\mathcal{L}(\lbrace n \rbrace) = \sum_{J} [n_J \log(1 + s_J/b_J) - s_J]$$

Exp. Log Likelihood ratio
$$\langle \mathcal{L}(\{n\}) \rangle_{\mathrm{SB}} = \sum_{J} \left[(s_J + b_J) \log \left(1 + \frac{s_J}{b_J} \right) - s_J \right]$$



Variance
$$\langle (\mathcal{L} - \langle \mathcal{L} \rangle_{\mathrm{SB}})^2 \rangle_{\mathrm{SB}} = \sum_J \left(b_j + s_J(M_{\mathrm{top}}) \right) \left[\log \left(1 + \frac{s_J(m)}{b_J} \right) \right]^2$$
 plotted $\mathcal{L} \pm \left[\left\langle (\mathcal{L} - \langle \mathcal{L} \rangle_{\mathrm{SB}})^2 \right\rangle_{\mathrm{SB}} \right]^{1/2}$

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