

BOOSTING $HH \rightarrow (b\bar{b})(b\bar{b})$ WITH MULTIVARIATE TECHNIQUES

BEHR, BORTOLETTO, FROST, ISSEVER, NH, ROJO [1512.08928]

Nathan Hartland
University of Oxford



Università degli Studi di Torino
Torino, 26/05/16

HIGGS PHYSICS AT THE LHC

ATLAS Prelim.

$m_H = 125.36 \text{ GeV}$

Phys. Rev. D 90, 112015 (2014)

$H \rightarrow \gamma\gamma$

$\mu = 1.17^{+0.27}_{-0.27}$

arXiv:1408.5191

$H \rightarrow ZZ^* \rightarrow 4l$

$\mu = 1.44^{+0.40}_{-0.33}$

arXiv:1412.2641

$H \rightarrow WW^* \rightarrow l\nu l\nu$

$\mu = 1.09^{+0.23}_{-0.21}$

arXiv:1409.6212

$W, Z H \rightarrow b\bar{b}$

$\mu = 0.5^{+0.4}_{-0.4}$

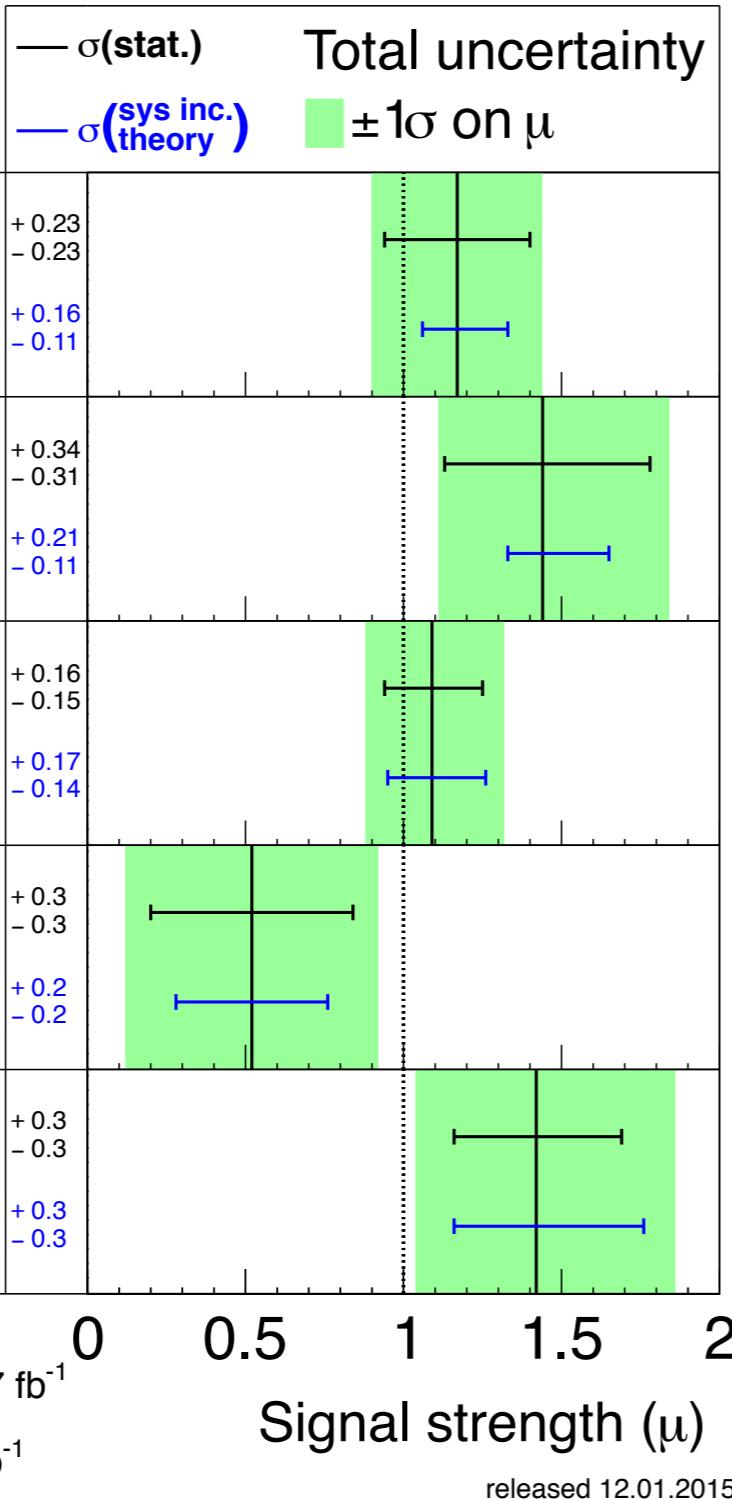
ATLAS-CONF-2014-061

$H \rightarrow \tau\tau$

$\mu = 1.4^{+0.4}_{-0.4}$

$\sqrt{s} = 7 \text{ TeV} \int L dt = 4.5-4.7 \text{ fb}^{-1}$

$\sqrt{s} = 8 \text{ TeV} \int L dt = 20.3 \text{ fb}^{-1}$



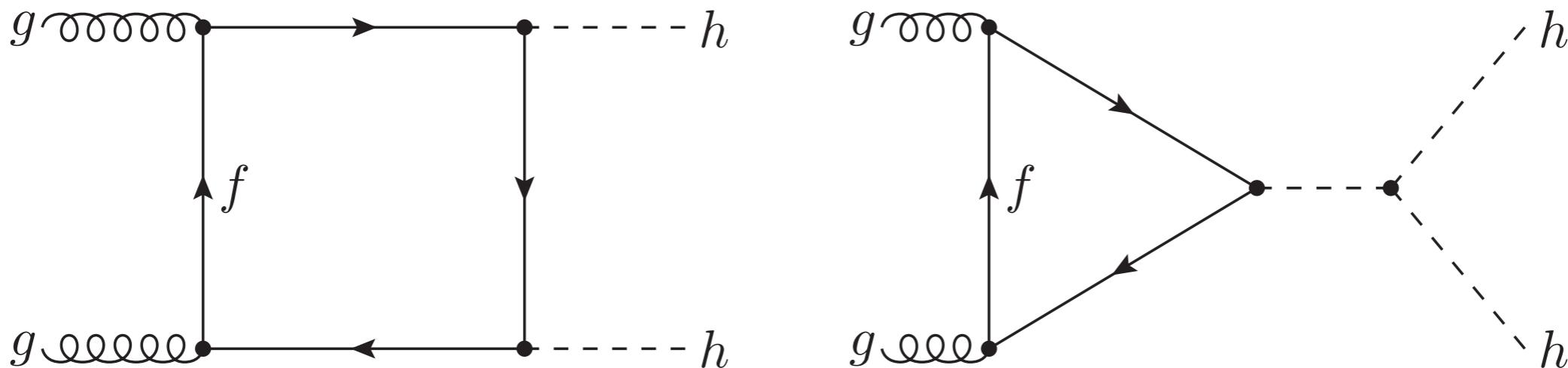
Resonance at $\sim 126 \text{ GeV}$ is pretty consistent with expectations for SM Higgs.

So far, we have only explored the **minimum** of the EWSB potential

$$V = \frac{1}{2} m_H^2 H^2 + v \lambda_3 H^3 + \frac{1}{4} \lambda_4 H^4$$

To understand the full potential, we need to measure (*at least*) double-Higgs production

DIHIGGS AT THE LHC



Clear difficulty: HH production cross-section at LHC is *tiny*

14TeV NNLO gluon fusion $\sigma_{HH} \simeq 40 \text{ fb}$ factor of $\sim 10^3$ smaller than single H

Compounded by usual H reconstruction problems

Final State	BF
$bbbb$	33%
$bbWW$	13%
$bb\tau\tau$	3.5%
$WWWW$	5.3%

Pr. Mode	Cross-sec
ggF	40 fb
VBF	2.0 fb
hhtt	1.0 fb
W/Z hh	0.5 fb

DIHIGGS AT THE LHC

Several LHC feasibility studies in ggF

$$b\bar{b}\tau^+\tau^-$$

Dolan, Englert, Spannowsky [1206.5001]

Papaefstathiou, Yang, Zurita [1209.1489]

$$b\bar{b}W^+W^-$$

Baur, Plehn, Rainwater [hep-ph/0304015]

Barr, Dolan, Englert, Spannowsky [1309.6318]

Dolan, Englert, Spannowsky [1206.5001]

$$b\bar{b}\gamma\gamma$$

Baur, Plehn, Rainwater [hep-ph/0310056]

Barger, Everett, Jackson [1311.2931]

Lu, Chang, Cheung, Lee, [1505.00957]

$$b\bar{b}b\bar{b}$$

Baur, Plehn, Rainwater [hep-ph/0304015]

Dolan, Englert, Spannowsky [1206.5001]

Wardrope, Jansen, et al [1410.2794]

Ferreira de Lima et al [1404.7139]

General picture:

Data is king: 3000 fb⁻¹ HL-LHC)

Significances $\sim 1\text{-}3\sigma$ / channel

Discovery of HH will require
the **combination** of channels

$HH \rightarrow (b\bar{b})(b\bar{b})$ BACKGROUNDS

Fully hadronic (gg)Higgs channel: experimentally challenging on a lot of fronts

Primary challenge: Overwhelming QCD background

Signal	Cross-section
HH	$4.0 \cdot 10^{-2} \text{ pb}$
Background	Cross-section
ZH	$7.7 \cdot 10^{-1} \text{ pb}$
ttH	$4.6 \cdot 10^{-1} \text{ pb}$
bbH	$6.1 \cdot 10^{-1} \text{ pb}$

Background	Cross-section
$bbbb$	$1.8 \cdot 10^3 \text{ pb}$
$bbjj$	$3.5 \cdot 10^5 \text{ pb}$
$jjjj$	$5.8 \cdot 10^6 \text{ pb}$
$tt(bbjjjj)$	$3.5 \cdot 10^3 \text{ pb}$

Previous studies: significances of order $\sim 2\sigma$

(without PU and missing some backgrounds)

$HH \rightarrow (b\bar{b})(b\bar{b})$ BACKGROUNDS

Background	Cross-section
bbbb	$1.8 \cdot 10^3$ pb
bbjj	$3.5 \cdot 10^5$ pb

Contributions from light jets usually discounted by b-tagging arguments. Assume (optimistically):

$$\epsilon_{\text{tag}} = 0.8 \quad \epsilon_{\text{mistag}} = 0.01$$

Then tagged 4b cross section $\sim \epsilon_{\text{tag}}^4 \cdot \sigma_{\text{bbbb}} = 7.3 \cdot 10^2$ pb

and tagged 2b2j cross section $\sim \epsilon_{\text{tag}}^2 \epsilon_{\text{mistag}}^2 \cdot \sigma_{\text{bbjj}} = 22$ pb

Consider instead the fraction of events containing n b-jets

$$(R = 0.4 \text{ a}k_T \quad p_T > 20 \text{ GeV})$$

	0 b-jet	1 b-jet	2 b-jet	3 b-jet	4 b-jet	ϵ_{sel}
bbbb	1%	8%	27%	44%	20%	8.4%
bbjj	9%	42%	49%	1%	0.1%	0.04%

$$\epsilon_{\text{sel}, \text{bbbb}} \cdot \sigma_{\text{bbbb}} \simeq 150 \text{ pb}$$

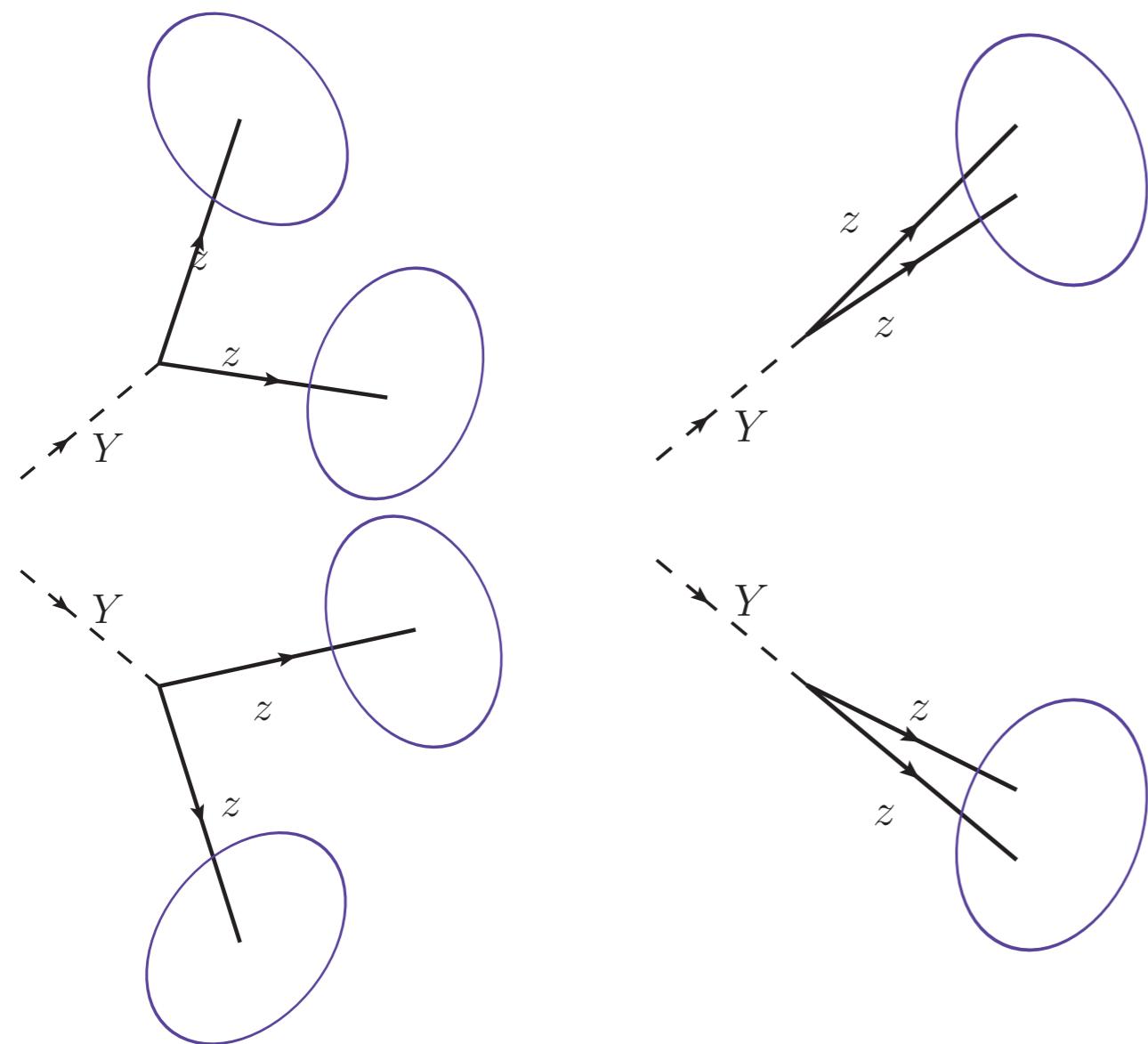
$$\epsilon_{\text{sel}, \text{bbjj}} \cdot \sigma_{\text{bbjj}} \simeq 140 \text{ pb}$$

Contributions e.g bbjj are not negligible w.r.t ‘irreducible’ 4b background

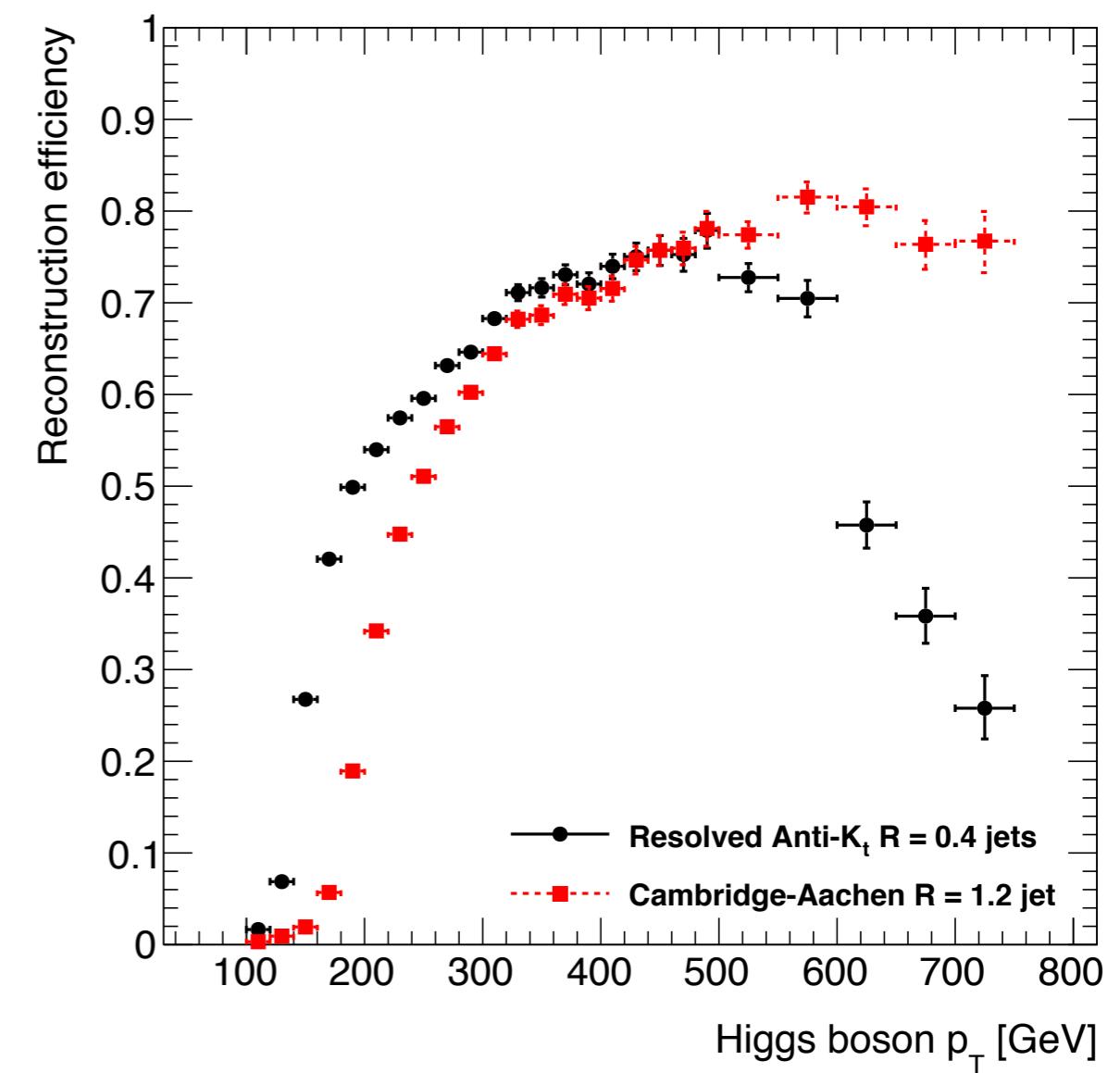
HH RECONSTRUCTION

$HH \rightarrow (b\bar{b})(b\bar{b})$ ANALYSIS TOPOLOGIES

- Resolved: Reconstruct H decay products in four separate (small-R) jets
- Boosted: Reconstruct H decay products in two (large-R) jets

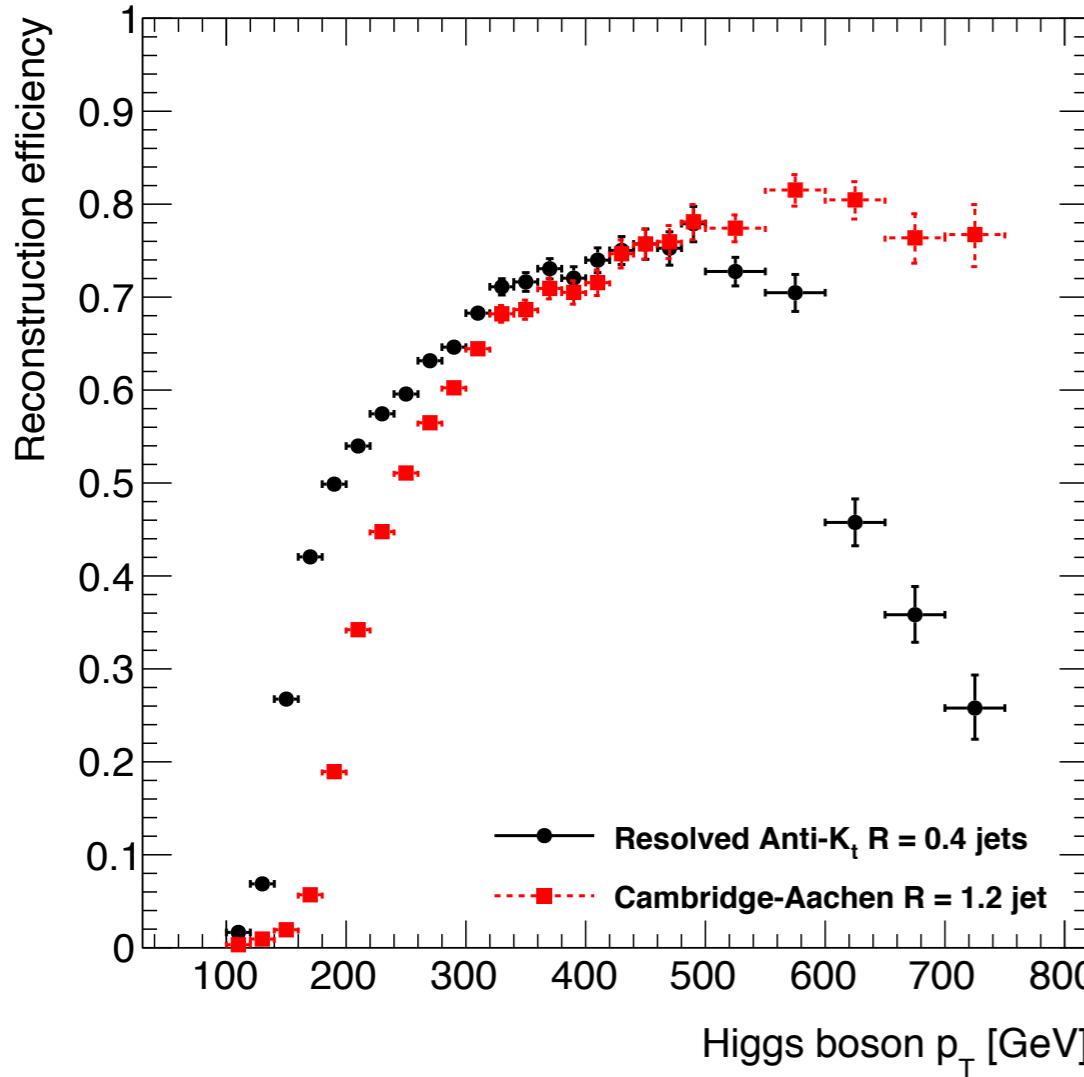


(Figs: Gouzevitch et al 1303.6636)



(Wardrobe et al 1410.2794)

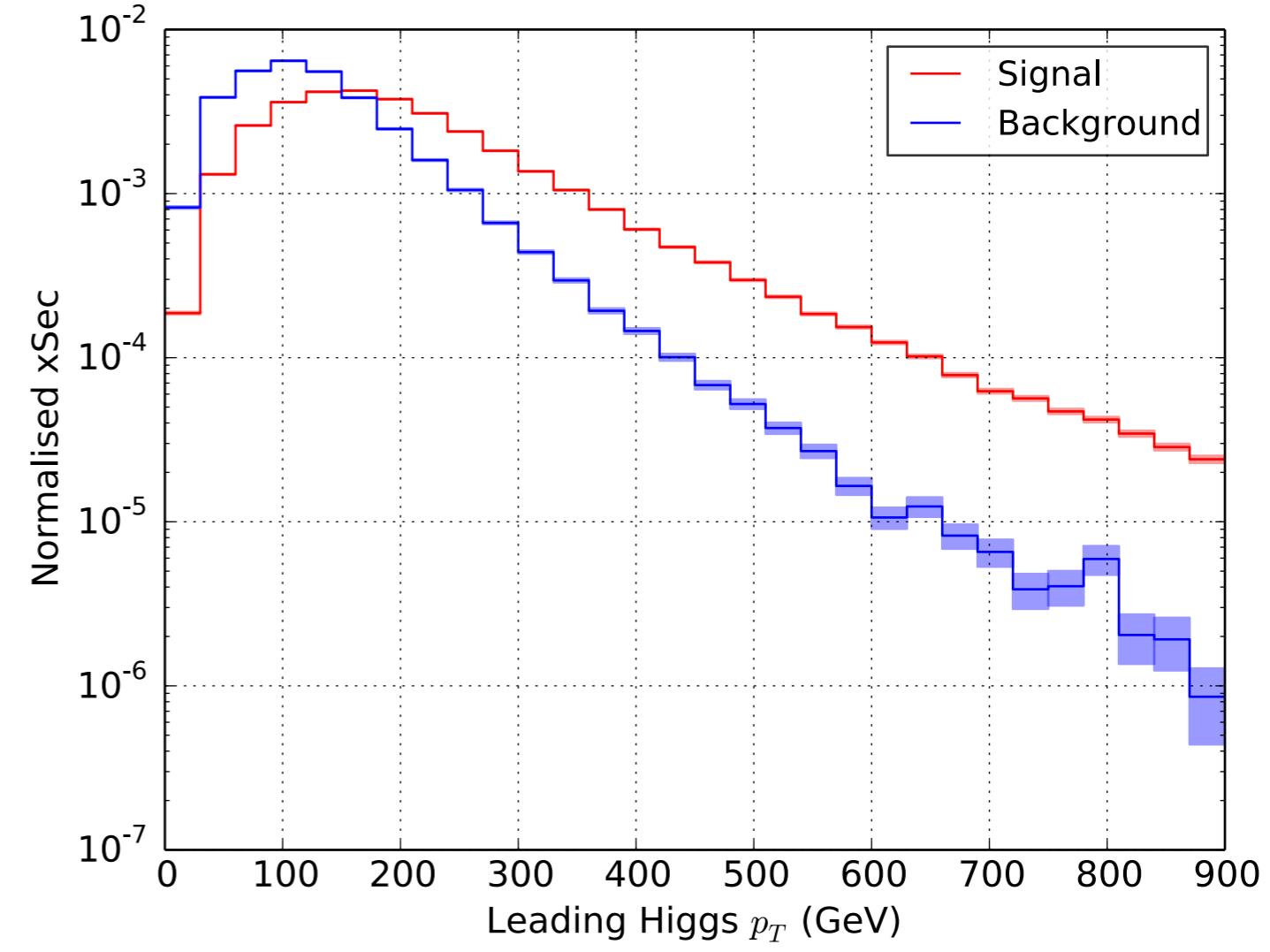
$HH \rightarrow (b\bar{b})(b\bar{b})$ ANALYSIS TOPOLOGIES



Resolved

- Captures bulk of HH cross-section
- Lower efficiency at high pT
- Challenging for triggers

(e.g UCL Study [1410.2794])

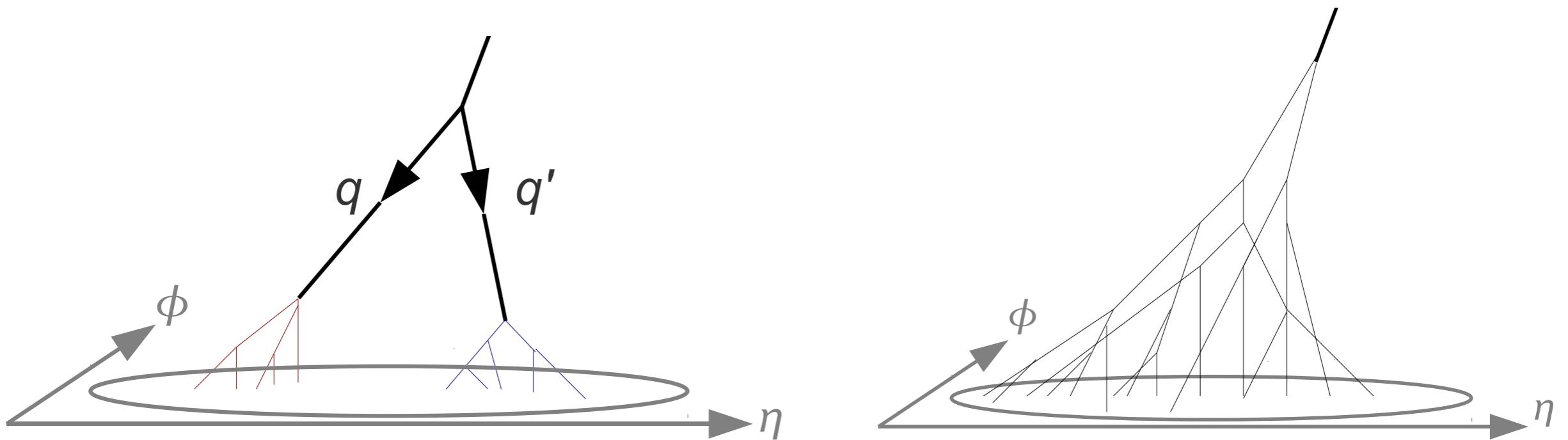


Boosted

- Better S/B but lower cross-sections
- Substructure tools available
- Greater sensitivity to pileup

(e.g Durham Study [1404.7139])

SUBSTRUCTURE FOR H->BB

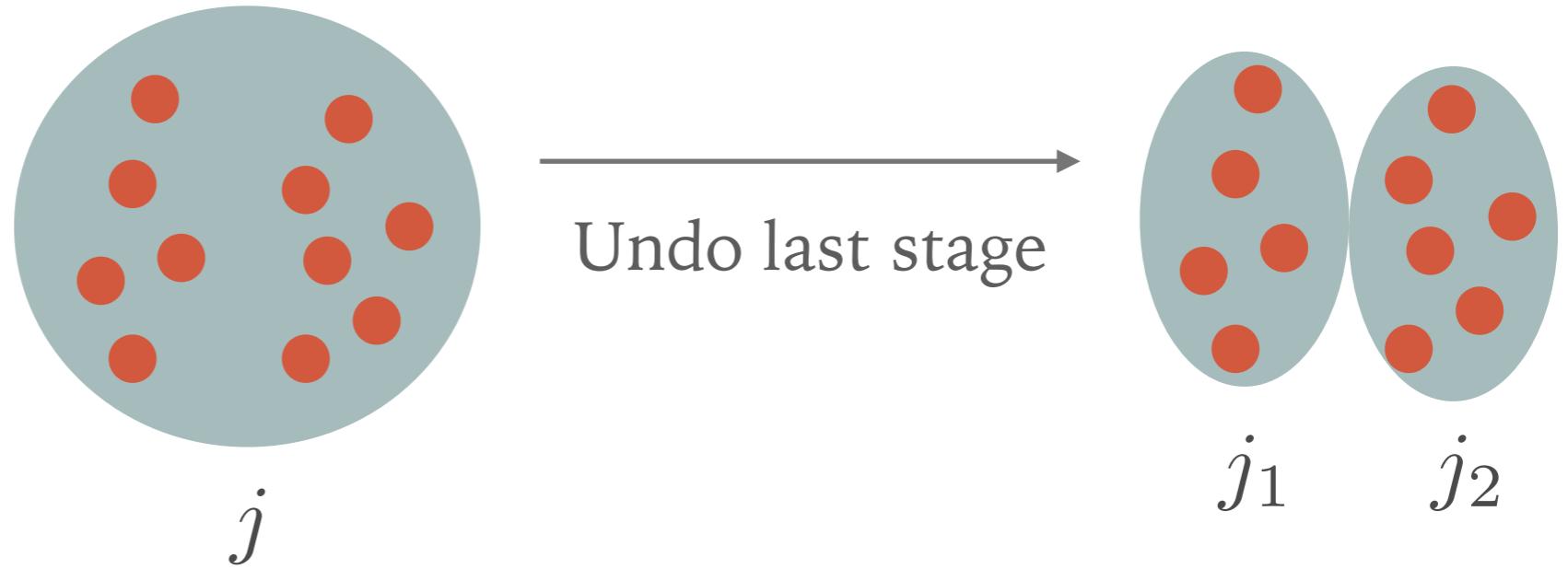


SUBSTRUCTURE - SPLITTING SCALES

[0802.2470], [hep-ph/0201098]

The kT algorithm clusters the two hardest subjets last

Cluster with large-R
kT algorithm

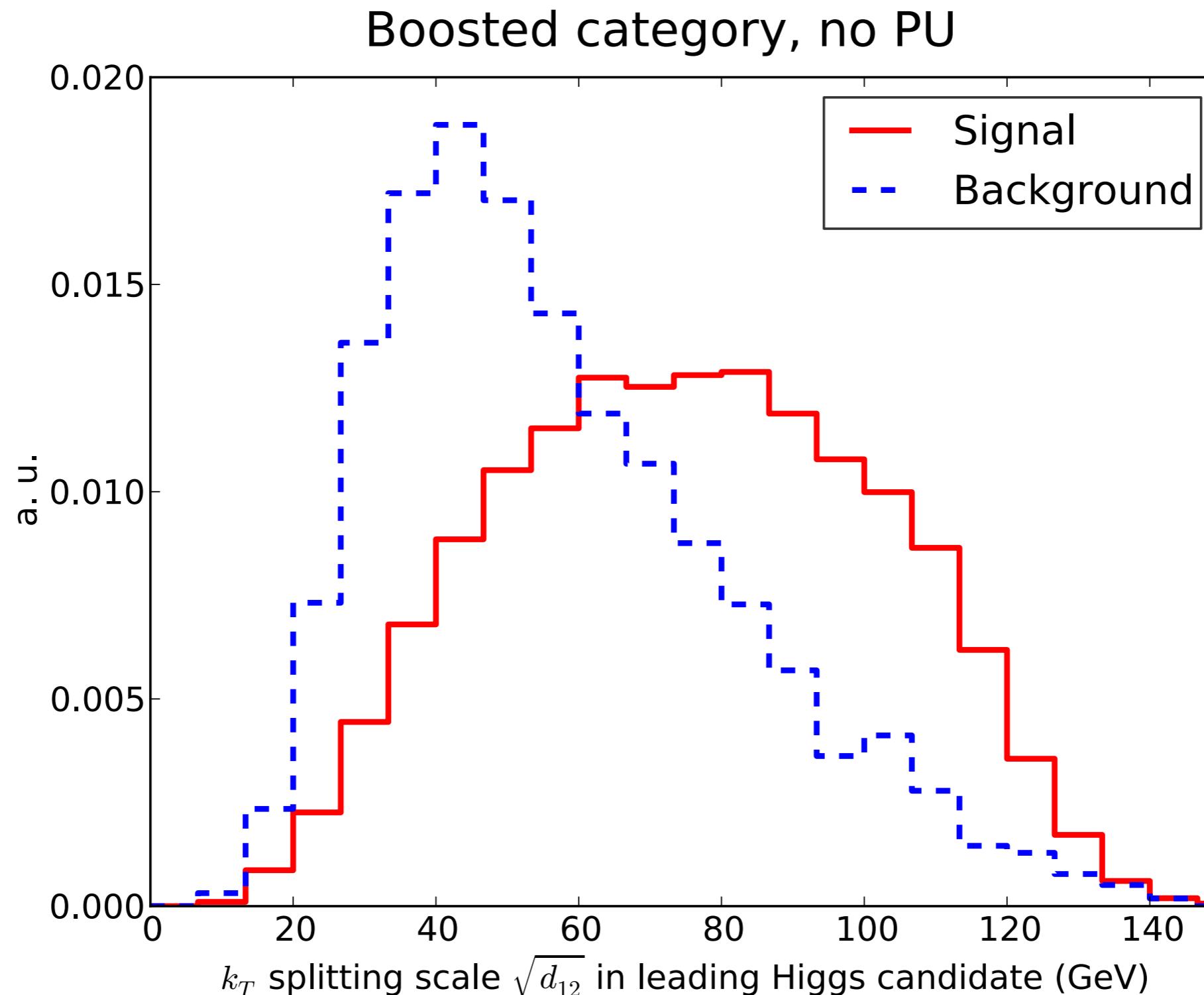


$$\text{The kT splitting scale } \sqrt{(d_{12})} = \min(p_{T,1}, p_{T,2}) \Delta R_{12}$$

Is the jet measure separating the two subjets at the final point of the combination

When separating H decay from QCD multi-jet background
H decay products are typically better separated than QCD

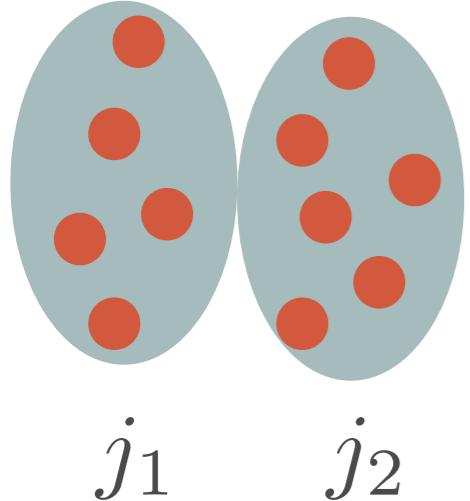
SUBSTRUCTURE - SPLITTING SCALES



SUBSTRUCTURE - N(SUB)-JETTINESS

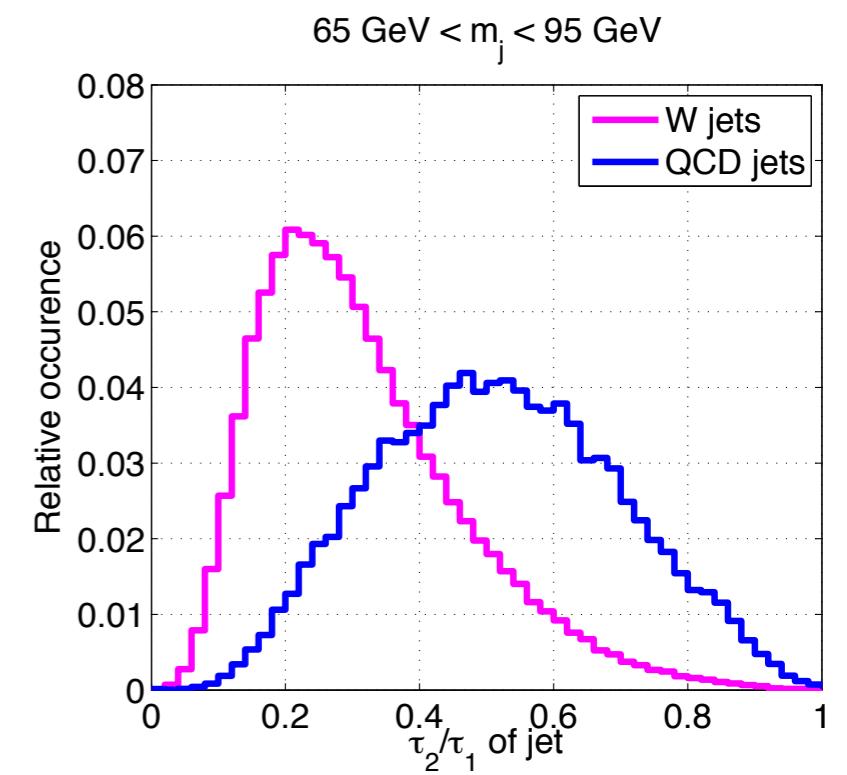
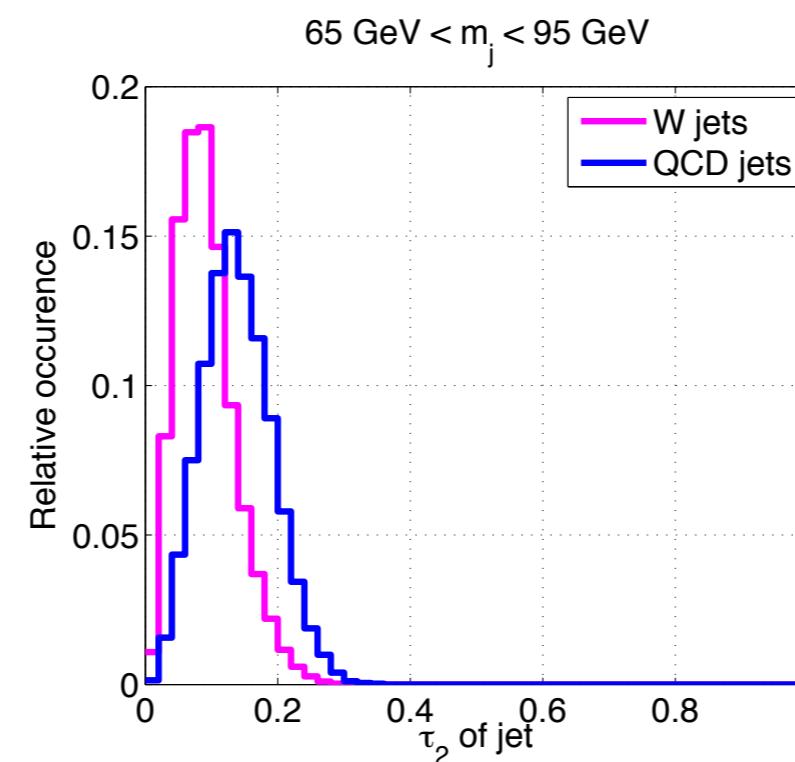
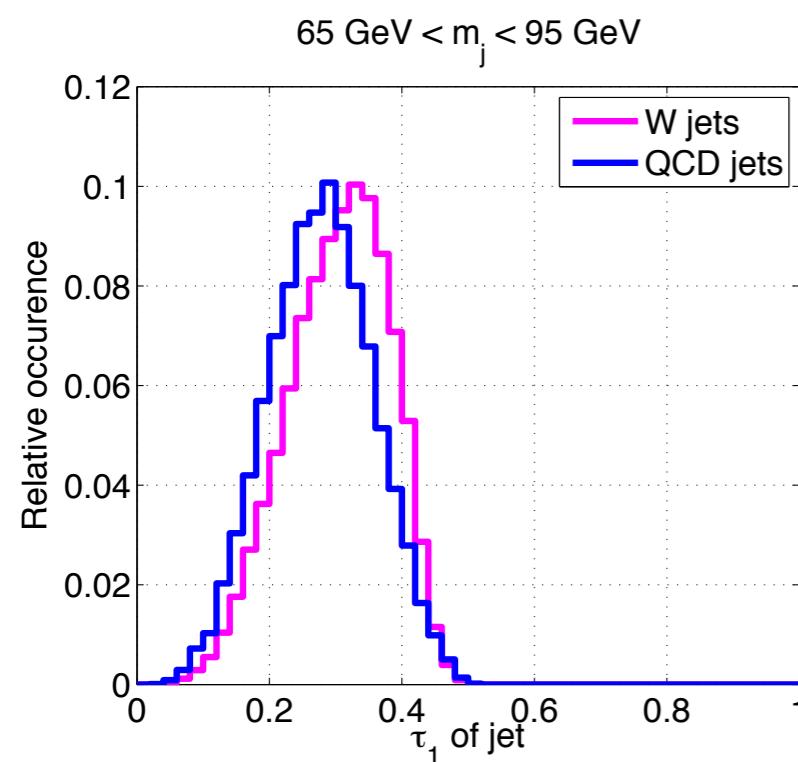
Thaler, Tilburg [1011.2268], [1108.2701]

Determine alignment of radiation along N subjet axes.



Recluster jet with exclusive kT algorithm to obtain N subjets, then compute how correlated jet radiation is along these axes

$$\tau_N \propto \sum_k p_{T,k} \min(\Delta R_{1,k}, \Delta R_{2,k} \dots \Delta R_{N,k})$$



(Figures [1011.2268])

SUBSTRUCTURE – ENERGY CORRELATIONS

Old idea, generalised by Larkoski, Salam, Thaler, [1305.0007]

Compute N-point Energy Correlation Function $\text{ECF}(N, \beta)$

e.g 2-point correlation $\text{ECF}(2, \beta) = \sum_{i < j \in J} p_{Ti} p_{Tj} (R_{ij})^\beta,$

Applied to a single jet

- For soft and collinear radiation: $\text{ECF}(2, \beta > 0) \rightarrow 0$

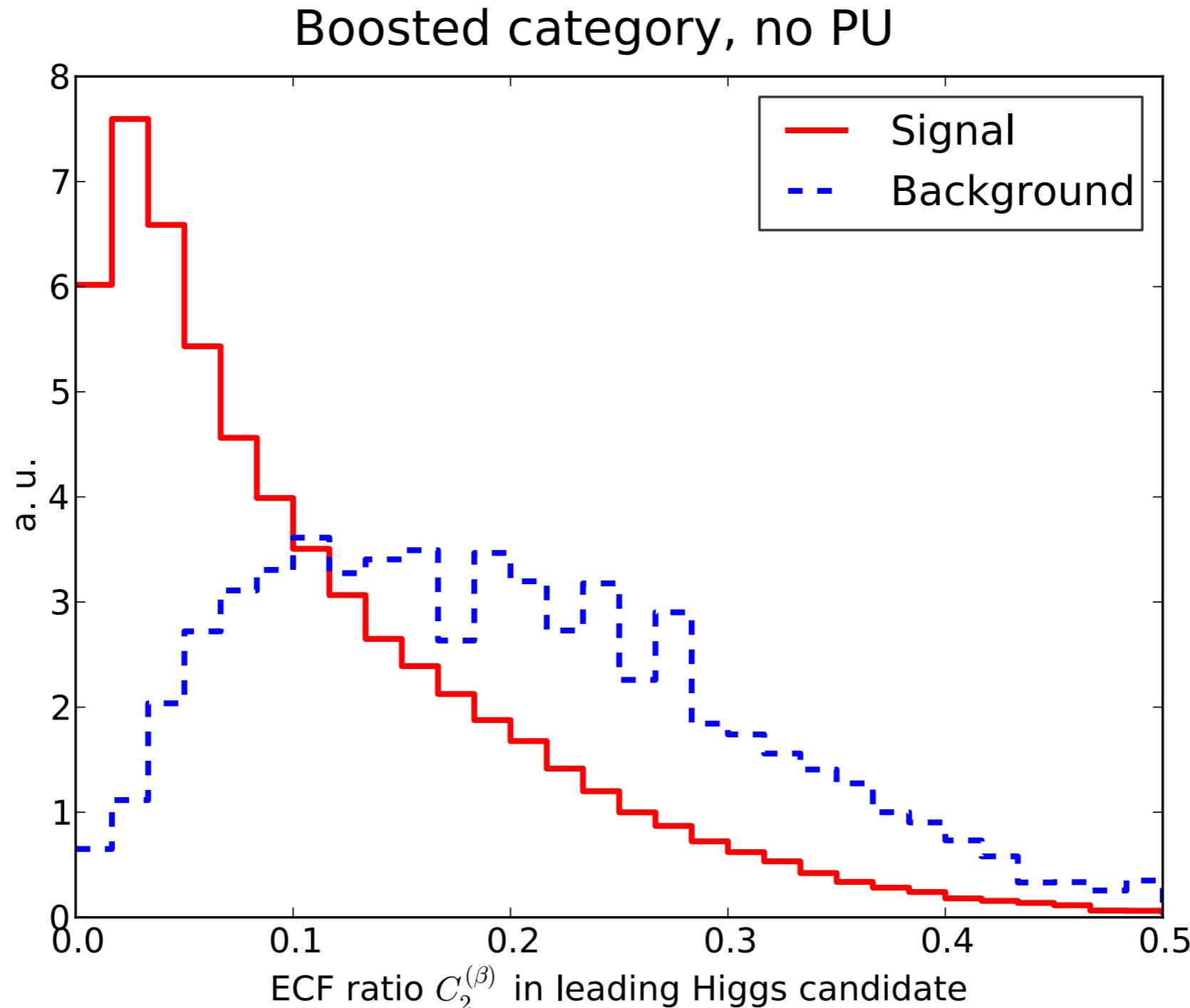
In order to get $\text{ECF}(2, \beta) \gg 0$ radiation must be well-separated and hard

Generally, N-point correlations are sensitive to (N-1) substructure

$$r_N^{(\beta)} \equiv \frac{\text{ECF}(N + 1, \beta)}{\text{ECF}(N, \beta)}$$

behaves similarly to N-subjettiness

SUBSTRUCTURE - ENERGY CORRELATIONS



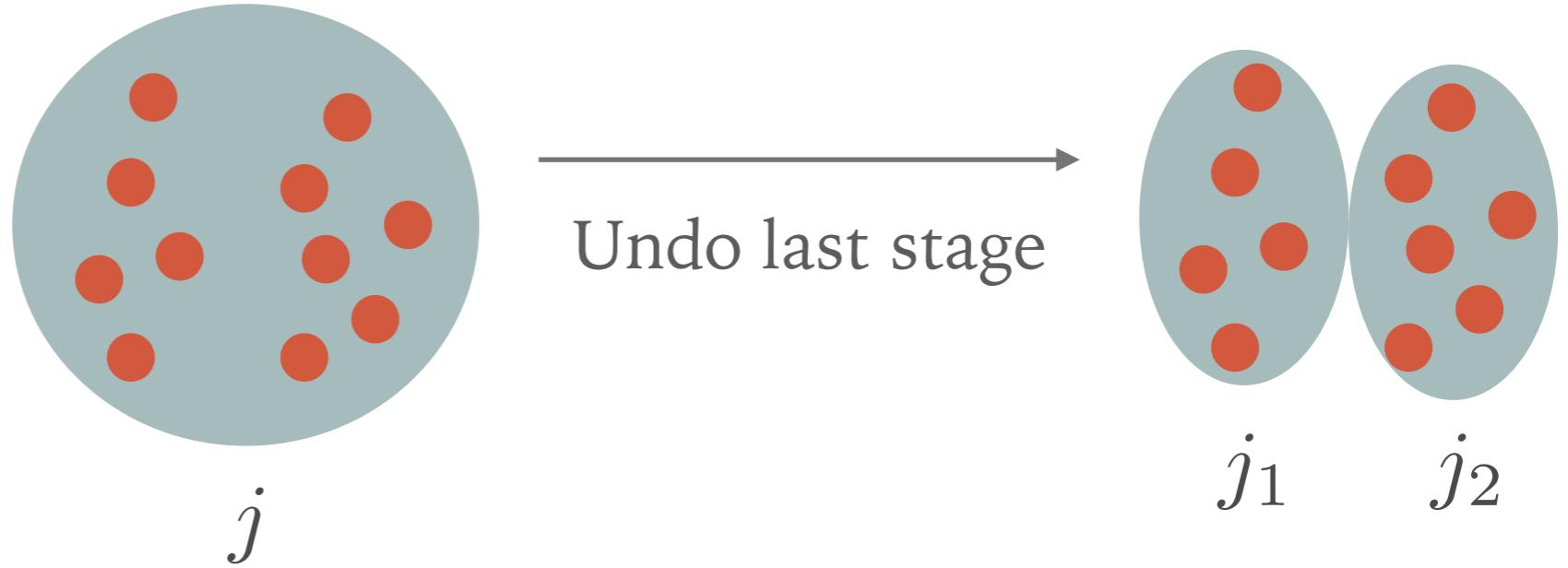
*Double-Ratios of ECFs (e.g. C_2, D_2) probe substructure like Nsubjettiness ratios
- but no need to define subjet axes!*

SUBSTRUCTURE – MASS DROP TAGS

Proposal of BDRS [0802.2470] for $H \rightarrow bb$

Identify fairly symmetric subjets with a significant drop in mass

Cluster with large-R
Cambridge-Aachen



Require $\max(m_{j1}, m_{j2}) < \mu m_j$ From symmetric subjets:

$$y = \min\left(p_T^{(j1)}, p_T^{(j2)}\right) \frac{\Delta R_{j1,j2}^2}{m_j^2} > y_{\text{cut}} \quad y \sim \frac{\min(z_{j1}, z_{j2})}{\max(z_{j1}, z_{j2})}$$

Provides good separation between heavy resonance decays and QCD

BOOSTING $HH \rightarrow (b\bar{b})(b\bar{b})$

OUR APPROACH

- Consider closely QCD multi-jet background (not just 4b)

Assuming relatively optimistic b-tagging parameters

$$\epsilon_b = 0.8, \quad \epsilon_c = 0.1, \quad \epsilon_j = 0.01$$

- Handle boosted and resolved analysis topologies

Ensure good coverage of final state phase space

- Use plenty of information on jet substructure

MD tags, Splitting scales, N-subjettiness, Energy correlations

- Investigate how multivariate analysis can boost significances

Keep cut-based analysis loose, try to make as much use as possible of MVA

- Assess robustness of results under the addition of pile-up (PU)

How feasible are these methods in a more realistic environment?

ANALYSIS DETAILS

Monte Carlo

Process	Generator	N_{evt}	$\sigma_{\text{LO}} \text{ (pb)}$	K -factor
$pp \rightarrow hh \rightarrow 4b$	MadGraph5_aMC@NLO	1M	$6.2 \cdot 10^{-3}$	2.4 (NNLO+NNLL [18, 19])
$pp \rightarrow bbbb$	SHERPA	3M	$1.1 \cdot 10^3$	1.6 (NLO [63])
$pp \rightarrow b\bar{b}jj$	SHERPA	3M	$2.7 \cdot 10^5$	1.3 (NLO [63])
$pp \rightarrow jjjj$	SHERPA	3M	$9.7 \cdot 10^6$	0.6 (NLO [77])
$pp \rightarrow t\bar{t} \rightarrow b\bar{b}jjjj$	SHERPA	3M	$2.5 \cdot 10^3$	1.4 (NNLO+NNLL [78])

Basic detector modelling

$$p_T \text{ Gaussian smear } p_T^{(i)} \rightarrow p_T^{(i)\prime} = (1 + r_i \cdot \sigma_E) p_T^{(i)}$$

$$\text{Calorimeter resolution } \Delta\eta \times \Delta\phi = 0.1 \times 0.1$$

‘Small-R jet’

b-tagged $aKT R=0.4$ jet
 $p_T > 40 \text{ GeV}, |\eta| < 2.5$

‘Large-R jet’

double b-tagged $aKT R=1.0$ jet
 $p_T > 200 \text{ GeV}, |\eta| < 2.5$
Require BDRS mass-drop tag

Higgs candidate (di-)jets required to have $|m - 125| \leq 40 \text{ GeV}$

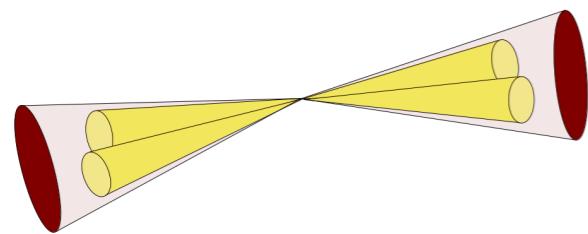
ANALYSIS DETAILS – SCALE INVARIANT RECONSTRUCTION

Need to make use of as much HH cross-section as possible

Combine all final state topologies Gouzevitch et al [1303.6636]

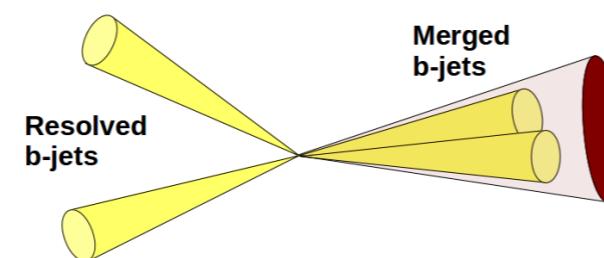
Boosted

Two Large-R jets



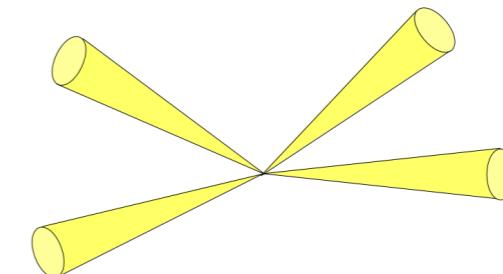
Intermediate

One Large-R jet, two Small-R jets



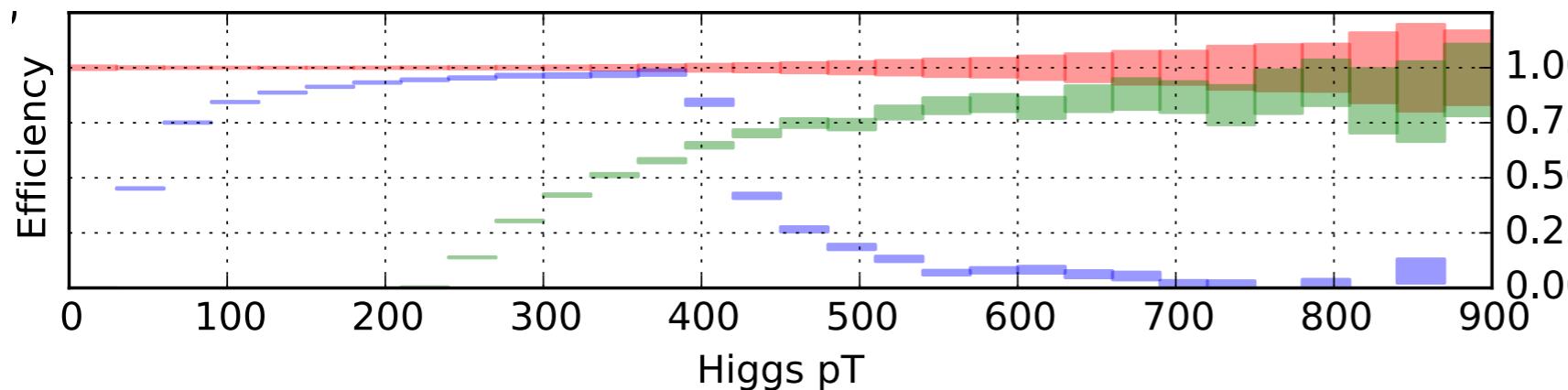
Resolved

Four Small-R jets



Prioritise selection Boosted-Intermediate-Resolved

Combined analysis covers wide range of Higgs pT



*Blue: Resolved efficiency
Green: Boosted efficiency*

PILEUP SIMULATION

PU may damage a measurement by (primarily)

- The introduction of new jets in reconstruction
- Distortion of jet masses and substructure

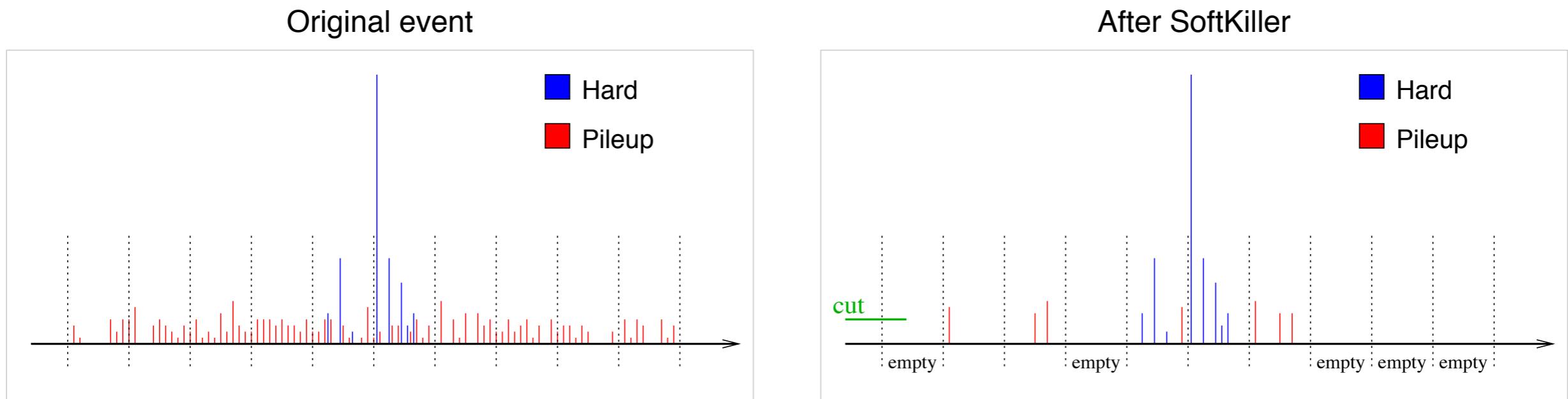
PU management will be a significant issue - particularly in boosted analyses.

-Introduce an additional 80 PU vertices per hard event.

PU is managed by a combination of

- SoftKiller subtraction at the event level

[Cacciari et al 1407.0408]



- Divide event into patches of $\Delta\eta \times \Delta\phi = a$
- Remove softest radiation until half of these patches are empty

PILEUP SIMULATION

PU may damage a measurement by (primarily)

- The introduction of new jets in reconstruction
- Distortion of jet masses and substructure

PU management will be a significant issue - particularly in boosted analyses.

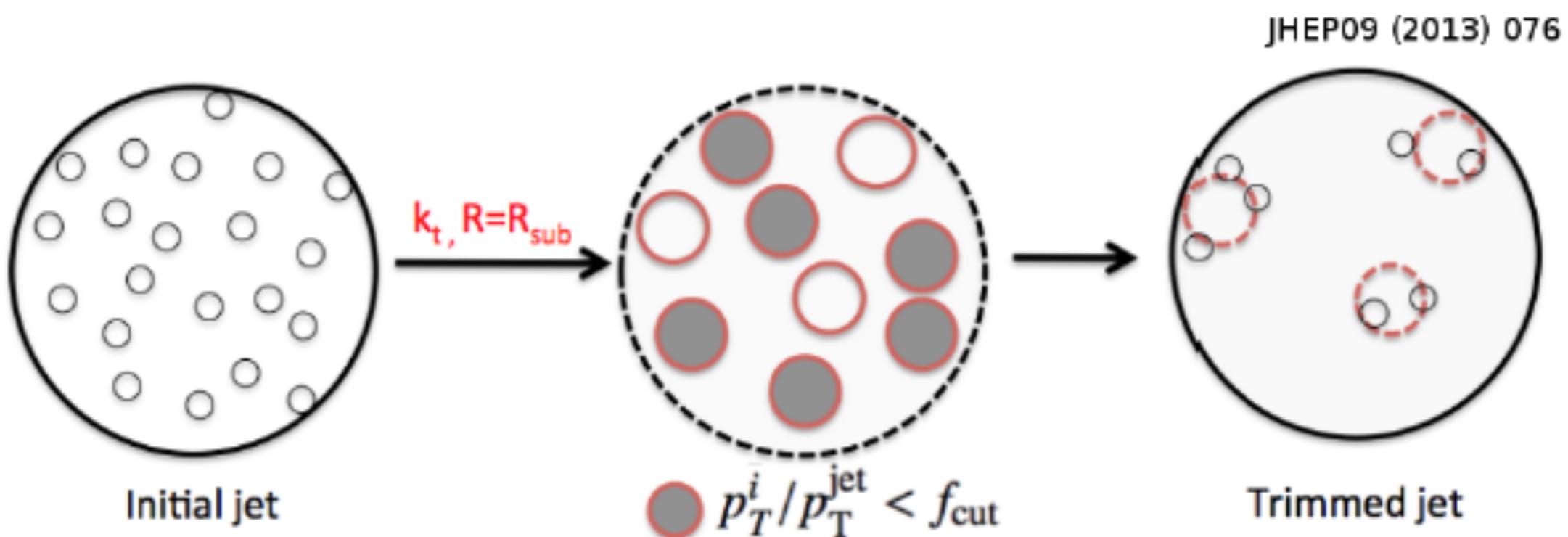
-Introduce an additional 80 PU vertices per hard event.

PU is managed by a combination of

- SoftKiller subtraction at the event level
- Large-R jets trimmed

[Cacciari et al 1407.0408]

[Krohn et al 0912.1342]



ANALYSIS RESULTS

ANALYSIS RESULTS

Results after cut-based analysis + PU80

	S/sqrt(B)	S/B
Boosted	0.6	9.0E-04
Intermediate	0.3	3.4E-04
Resolved	0.4	6.5E-05

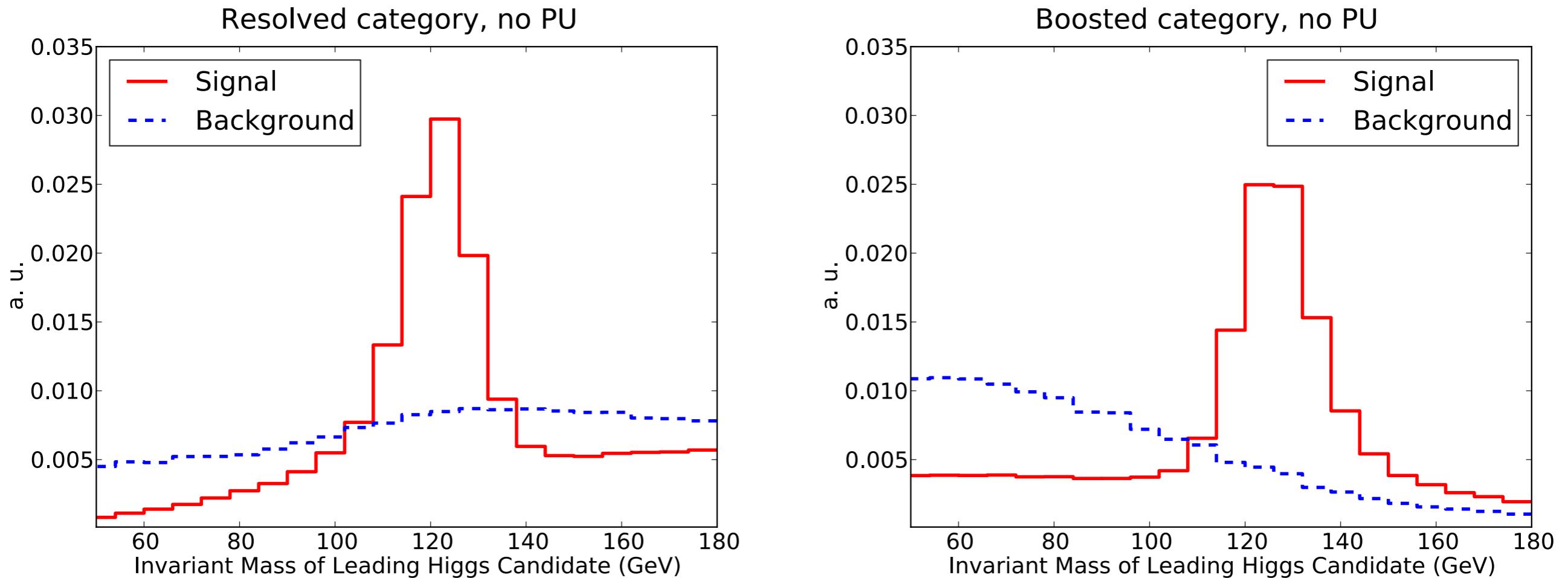
	HL-LHC	NSIG	NBKG
Boosted	410	4.5E+05	
Intermediate	260	7.7E+05	
Resolved	1800	2.7E+07	

Combined statistical significance before MVA $\sim 0.8\sigma$

Resolved	HH	QCD 4b	QCD 2b2j	QCD 4j	QCD ttbar
Reco (fb)	11	1.5E+05	3.0E+07	4.1E+08	2.6E+05
Final (fb)	0.6	3.5E+03	5.1E+03	3.1E+02	50

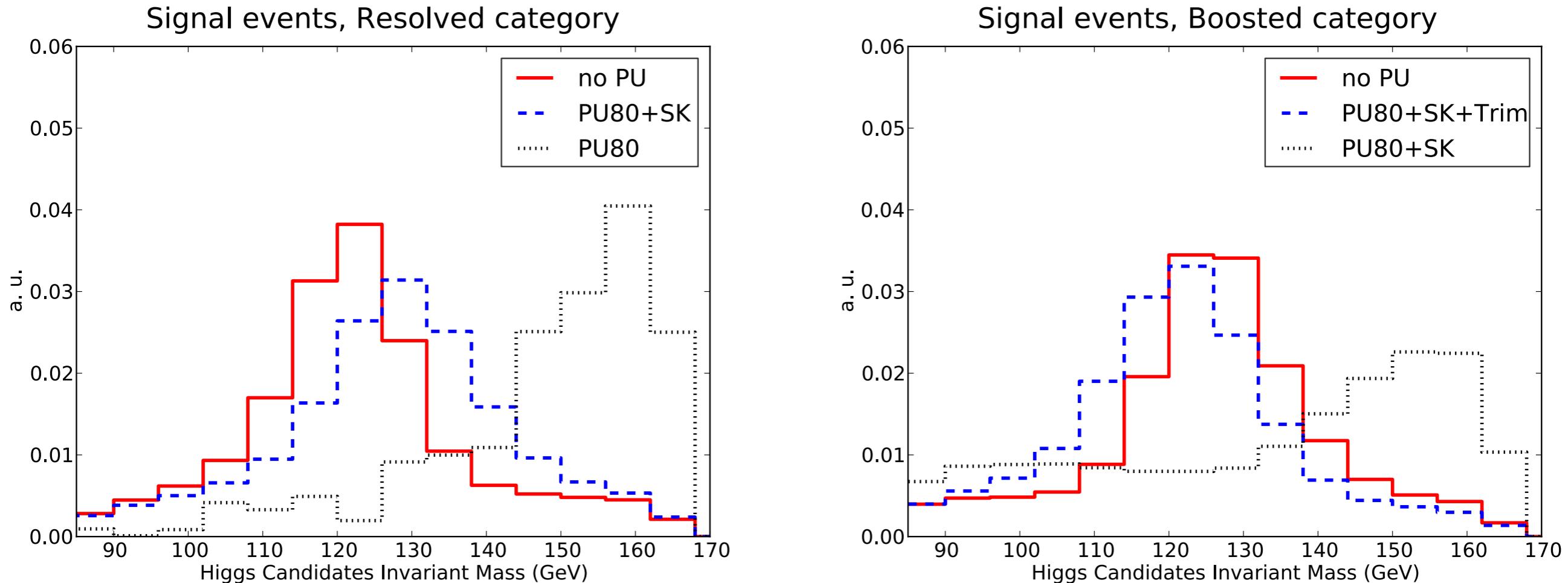
Boosted	HH	QCD 4b	QCD 2b2j	QCD 4j	QCD ttbar
Reco (fb)	3.5	1.0E+04	2.7E+06	3.8E+07	2.0E+04
Final (fb)	0.14	40	86	22	1.8

ANALYSIS RESULTS – PU AND JET MASSES



- *(di)-jet masses are naturally a vital component in any H measurement*
 - Can the effects of pile-up be successfully mitigated?

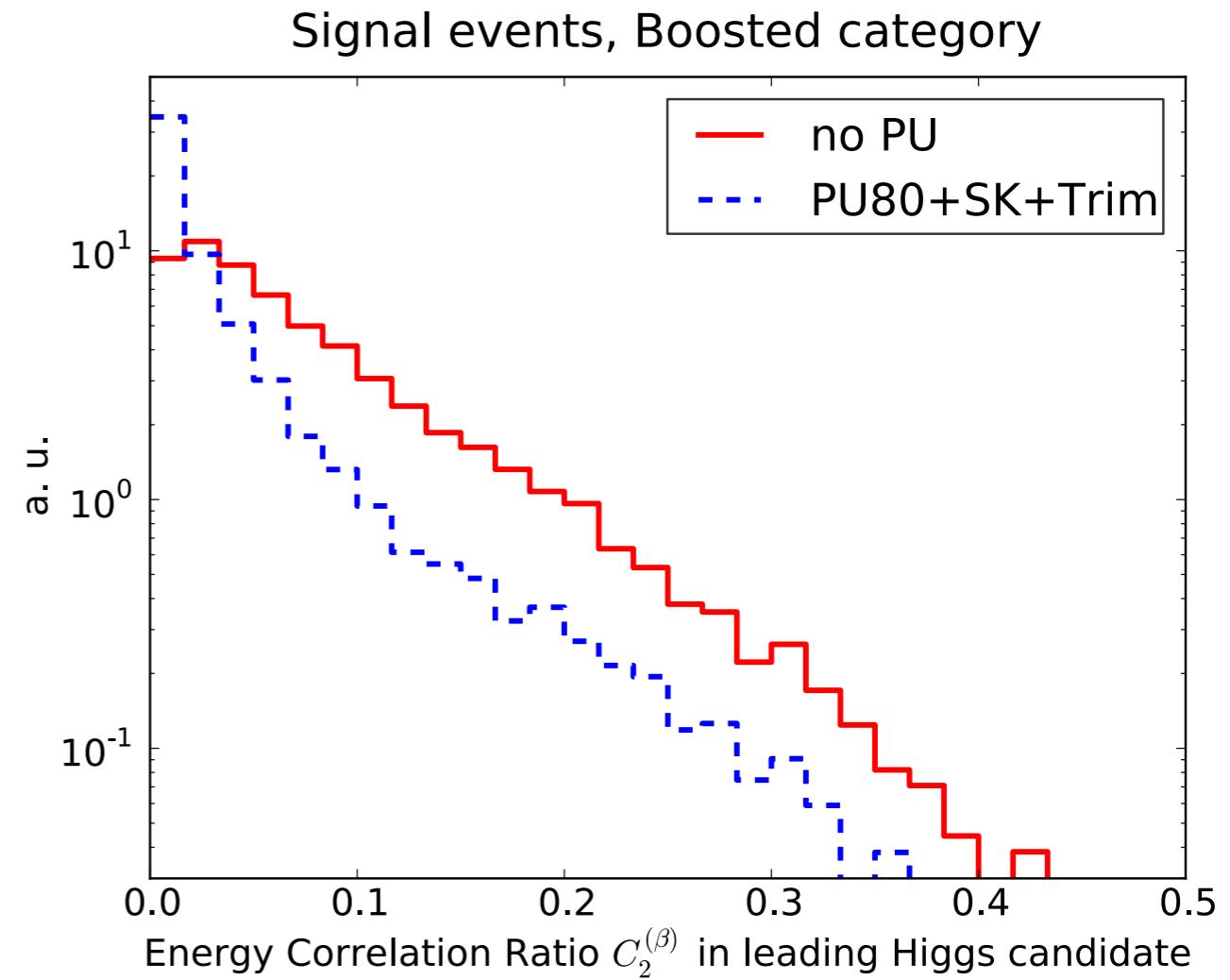
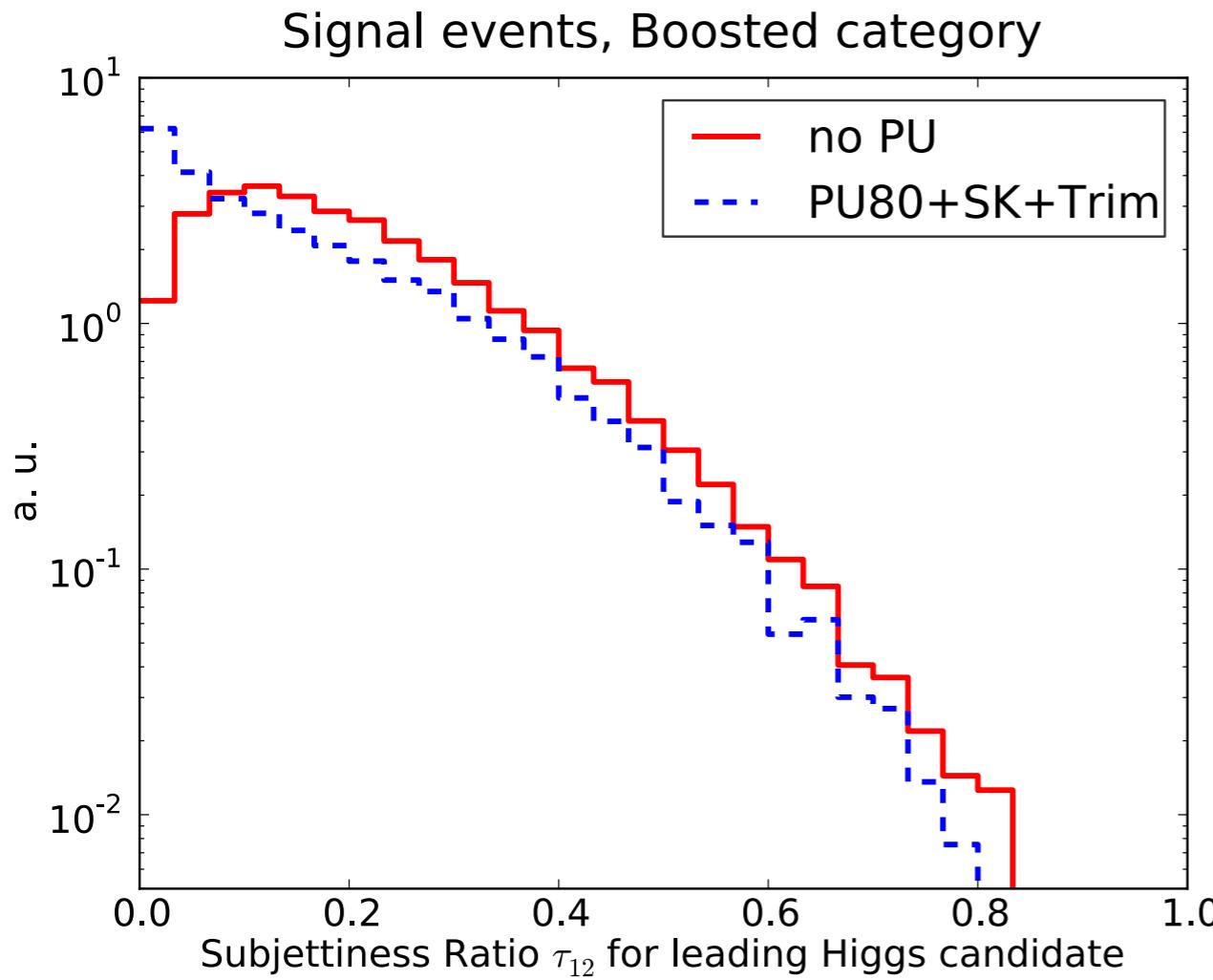
ANALYSIS RESULTS – PU AND JET MASSES



- *Resolved di-jet masses: SoftKiller appears to function well*
- *Boosted jet masses: SoftKiller itself isn't sufficient, some Grooming required*

ANALYSIS RESULTS – SUBSTRUCTURE AND PILEUP

- Substructure observables are often naturally sensitive to pile-up



With SK + Trimming for pile-up subtraction:

Principle features of substructure variables appear to be fairly robust

MULTIVARIATE

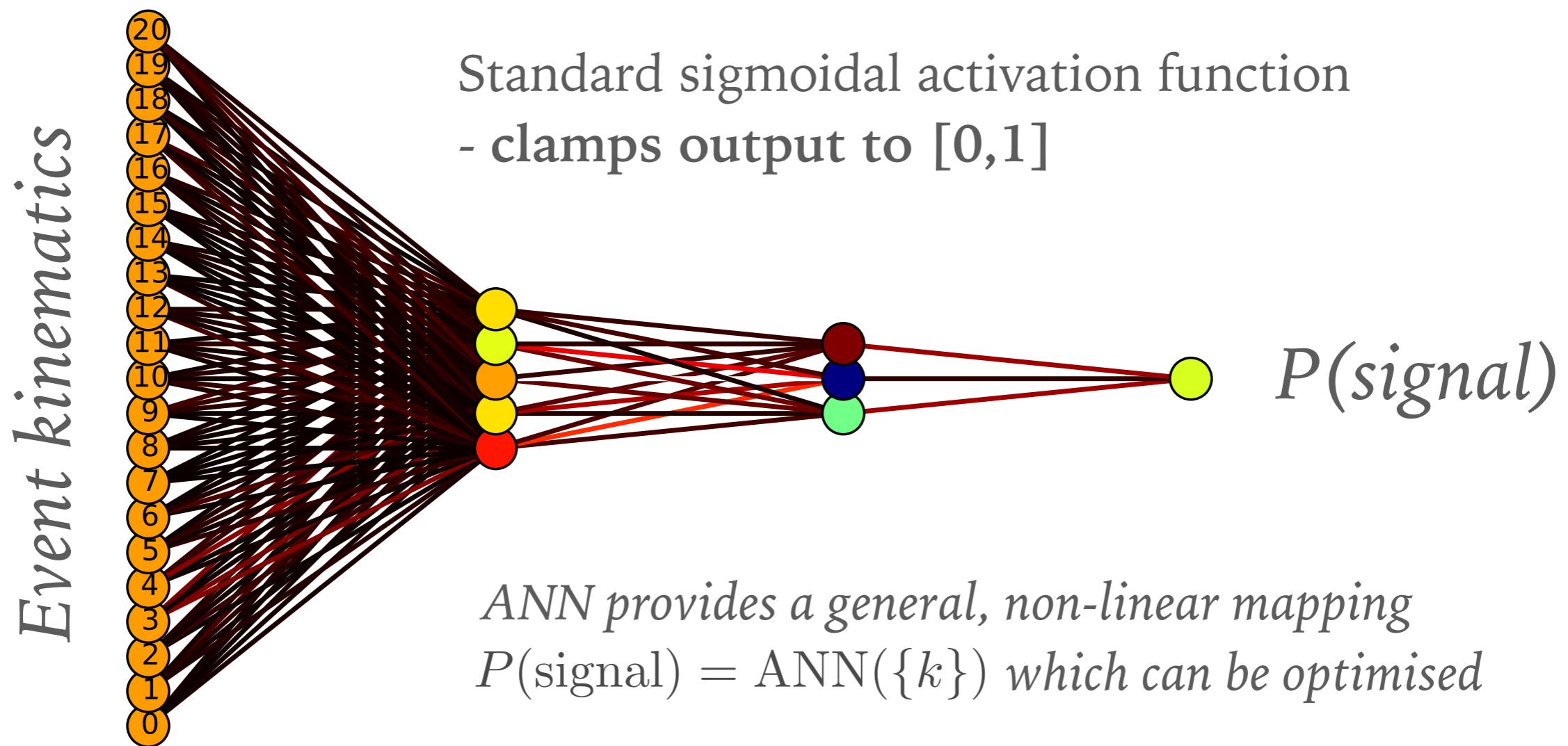
$HH \rightarrow (b\bar{b})(b\bar{b})$

MULTIVARIATE ANALYSIS – NEURAL NET CLASSIFIERS

Cuts so far are rather loose - how can we best make use of our information?

Classify events as signal/background with ANN

$N \times 5 \times 3 \times 1$ (Similar architecture to NNPDF)



MULTIVARIATE ANALYSIS – KINEMATIC INPUTS

Event kinematics used as input to the neural network

For each event

- p_{T,H_1} , p_{T,H_2} , $p_{T,HH}$
- m_{H_1} , m_{H_2} , m_{HH}
- ΔR_{HH} , $\Delta\eta_{HH}$, $\Delta\phi_{HH}$
- $p_{T,b\text{-jet}}$ for each b -(sub)jet

For each large-R jet

- k_T splitting scales $\sqrt{(d_{12})}$
- 2/1 - subjettiness τ_2/τ_1
- ECF double ratio $C_2^{(\beta)}$
- ECF double ratio $D_2^{(\beta)}$

Channel	#inputs
Boost	21
Inter.	17
Resol.	13

ANN untroubled by $\mathcal{O}(10)$ inputs:
Let the MVA decide which variables are
the most discriminating

MULTIVARIATE ANALYSIS – TRAINING

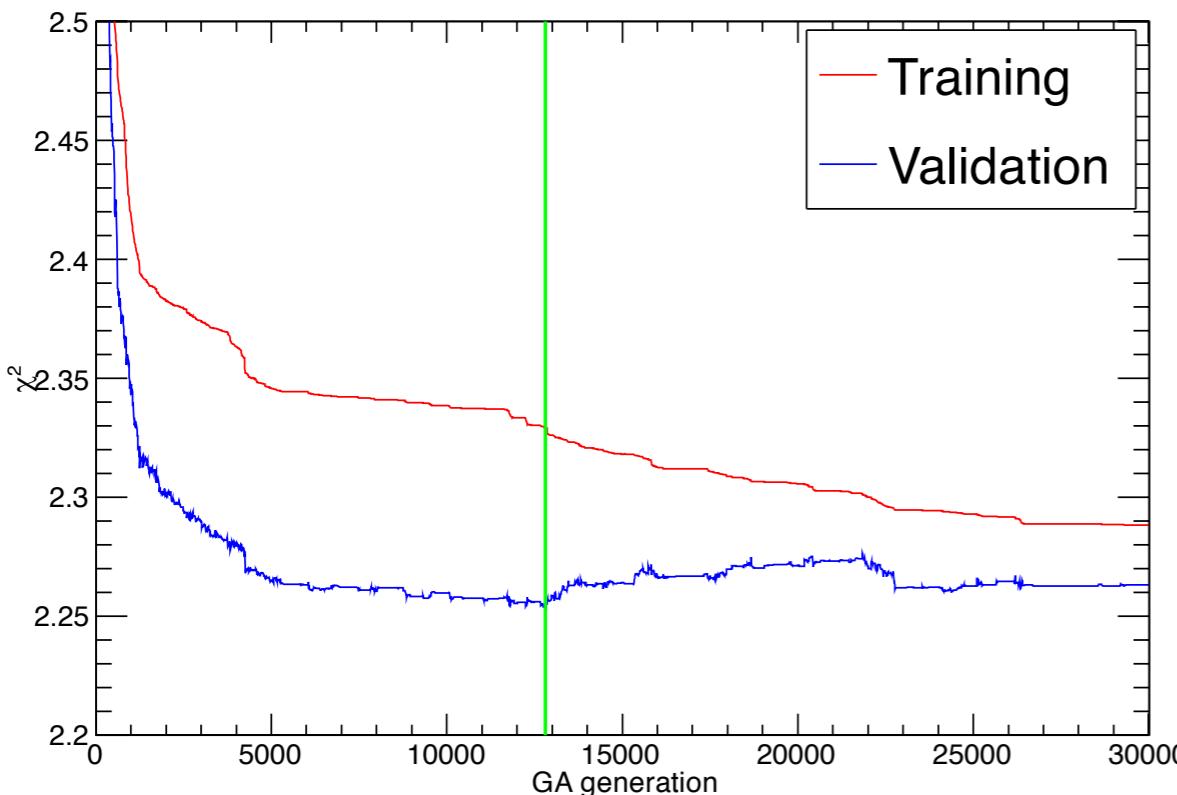
The neural network is trained upon Monte Carlo events

The optimisation consists of

minimising the Cross-Entropy

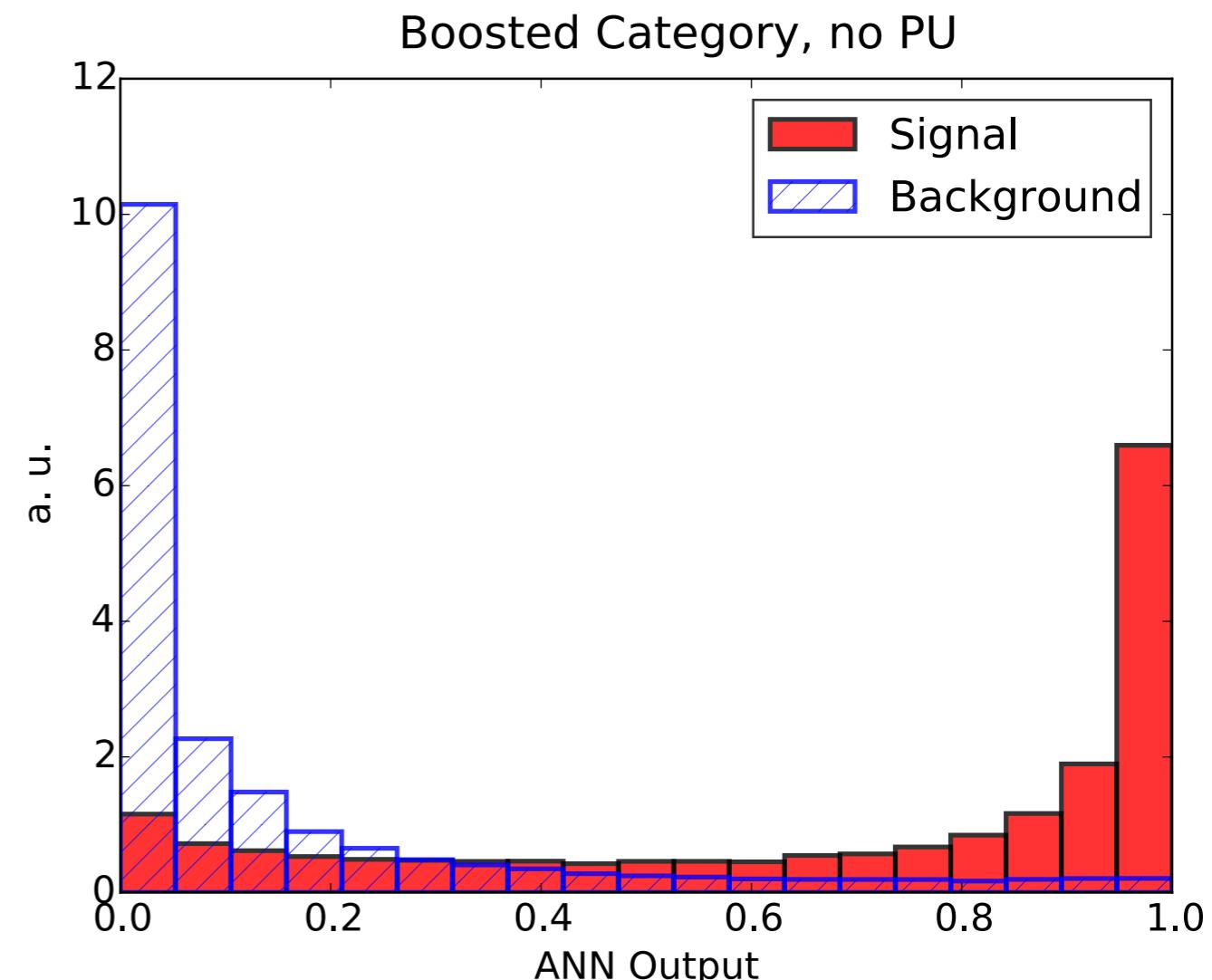
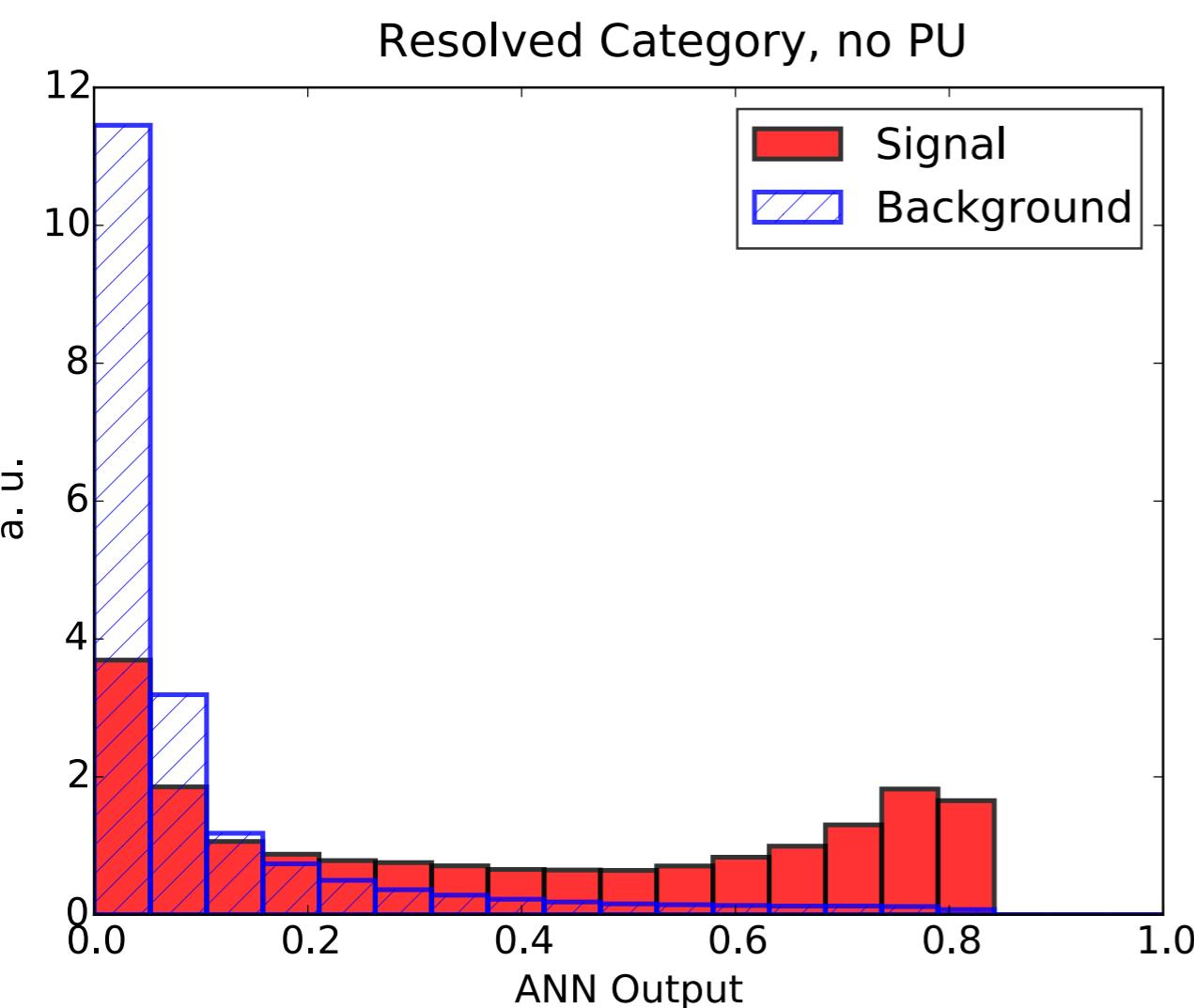
$$E(\{\omega\}) \equiv -\log \left(\prod_i^{N_{\text{ev}}} P(y'_i | \{k\}_i, \{\omega\}) \right)$$
$$= \sum_i^{N_{\text{ev}}} [y'_i \log y_i + (1 - y'_i) \log (1 - y_i)]$$

- The minimisation progresses through a basic genetic algorithm
- Over-fitting is prevented by ‘look-back’ cross-validation



- Split data into training and validation
- Minimise to training data
- Select GA iteration providing the best fit to the validation data

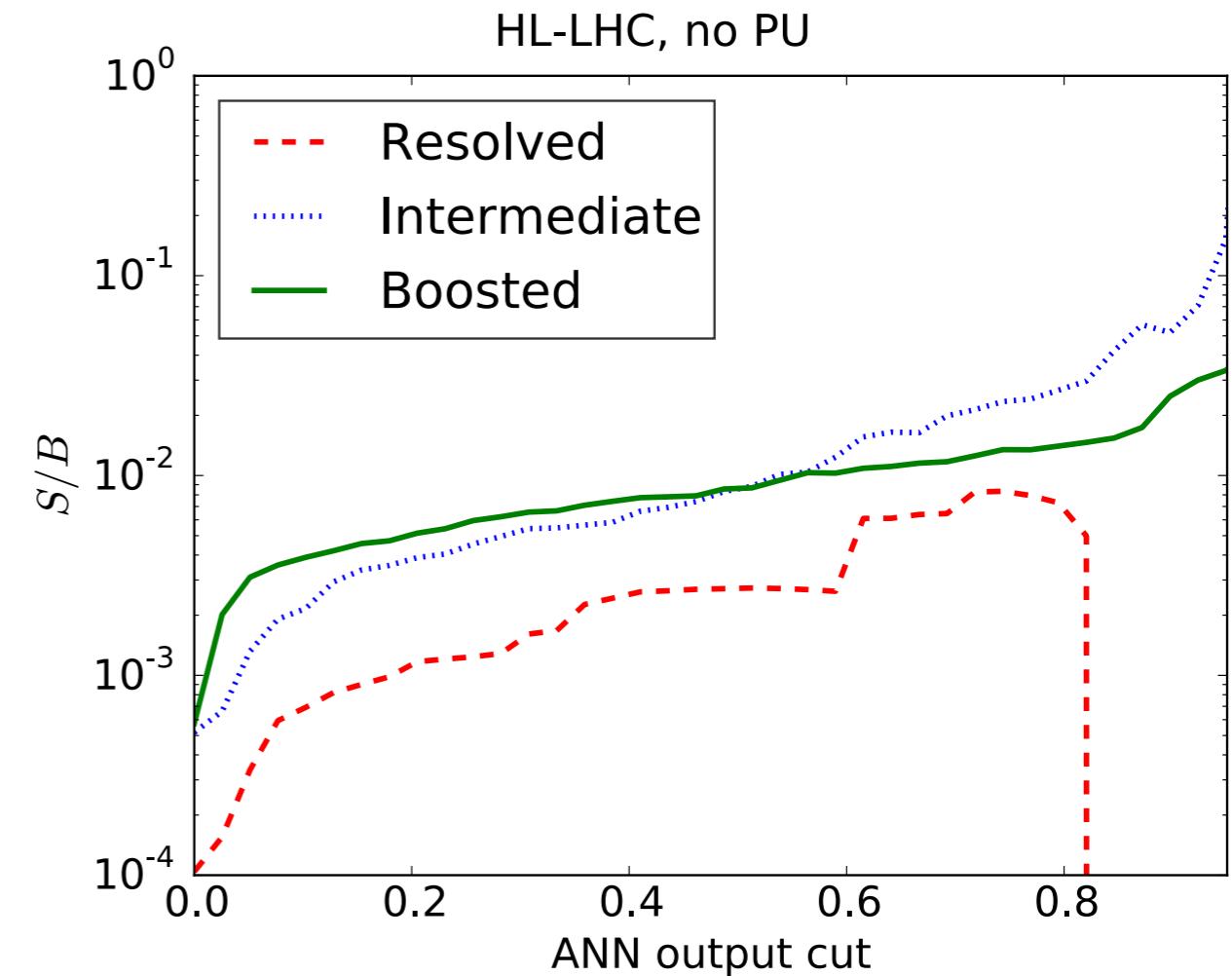
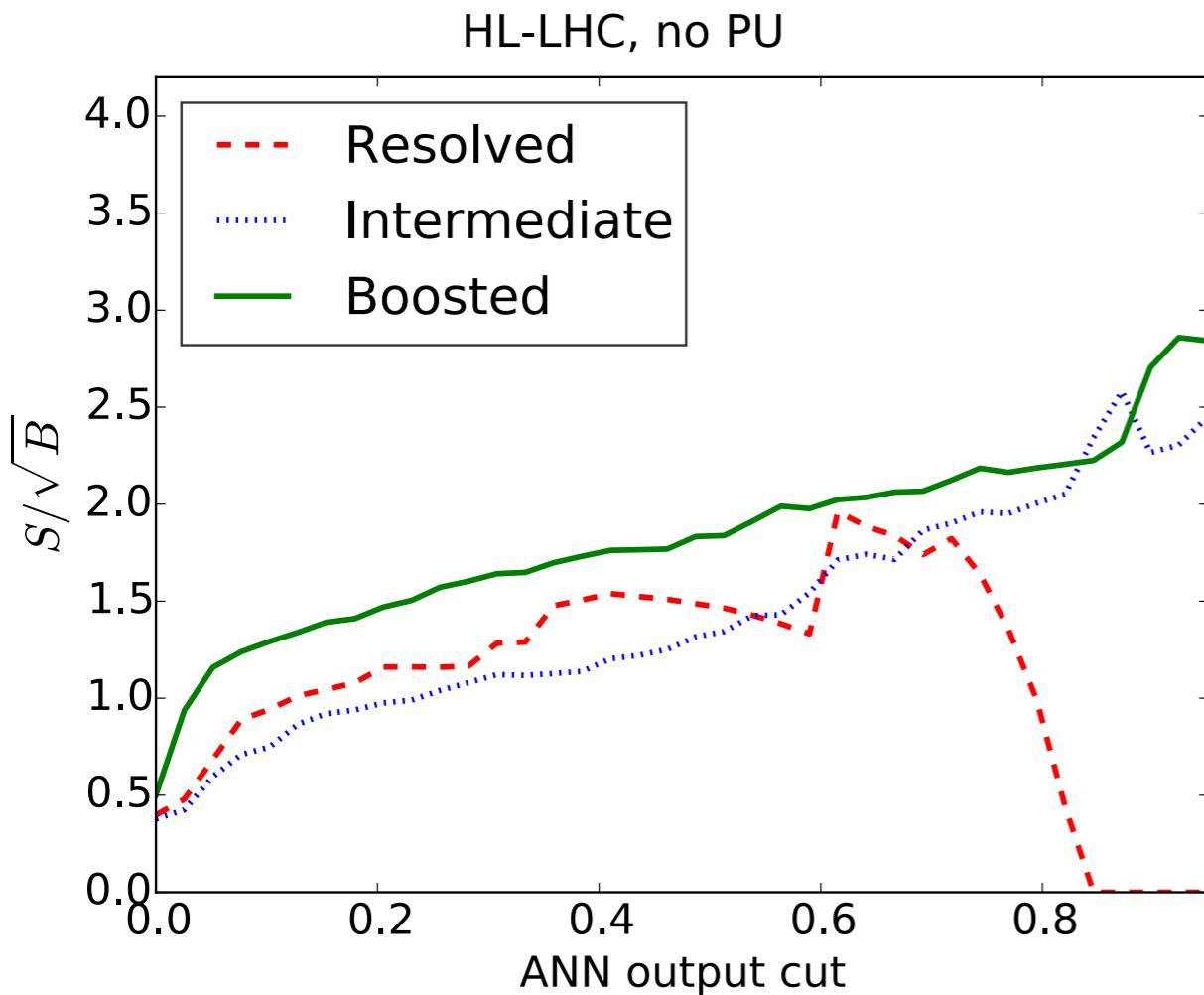
MULTIVARIATE ANALYSIS – RESULTS



Simple ANN training yields reasonable signal-background separation

MULTIVARIATE ANALYSIS – RESULTS

- Vary the cut in ANN output (no pileup case)

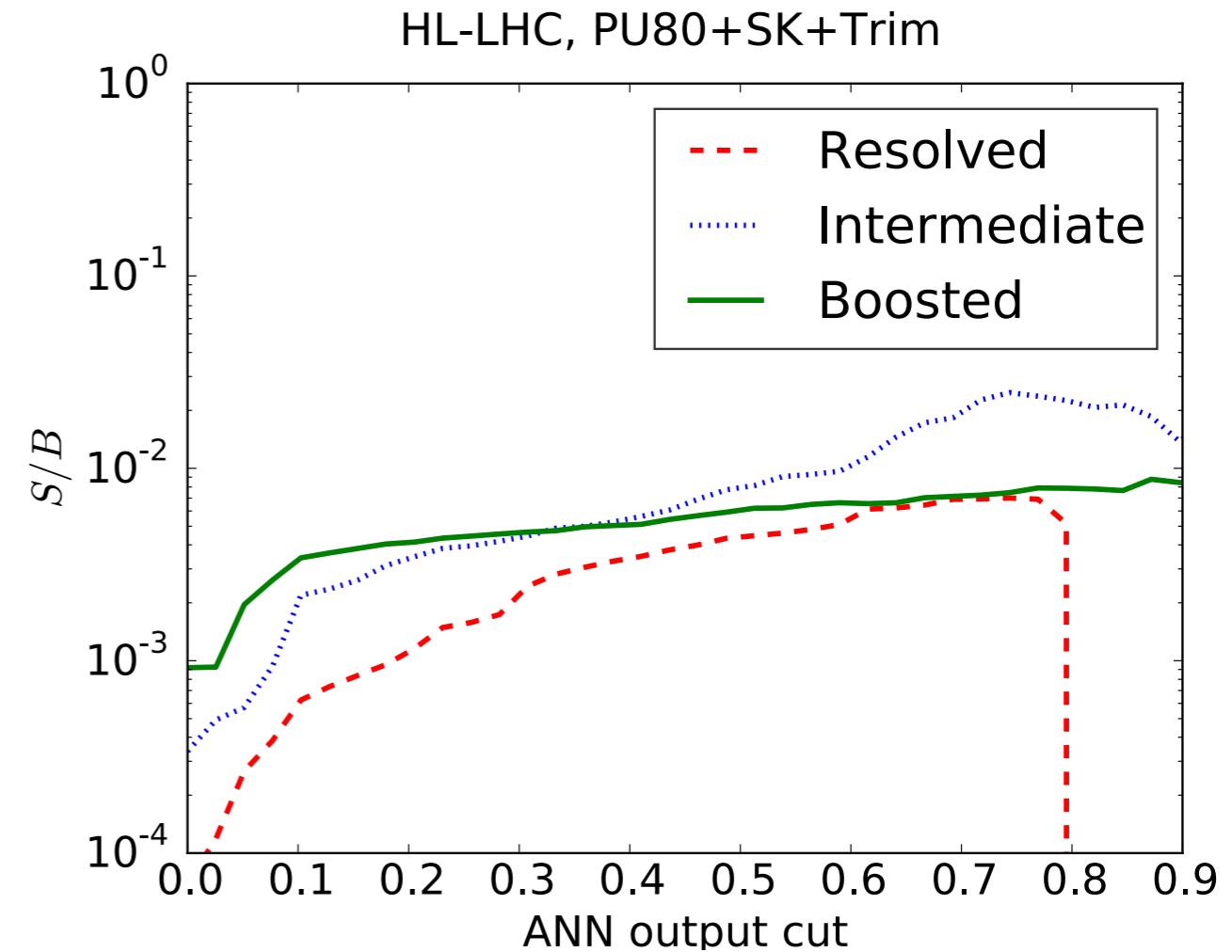
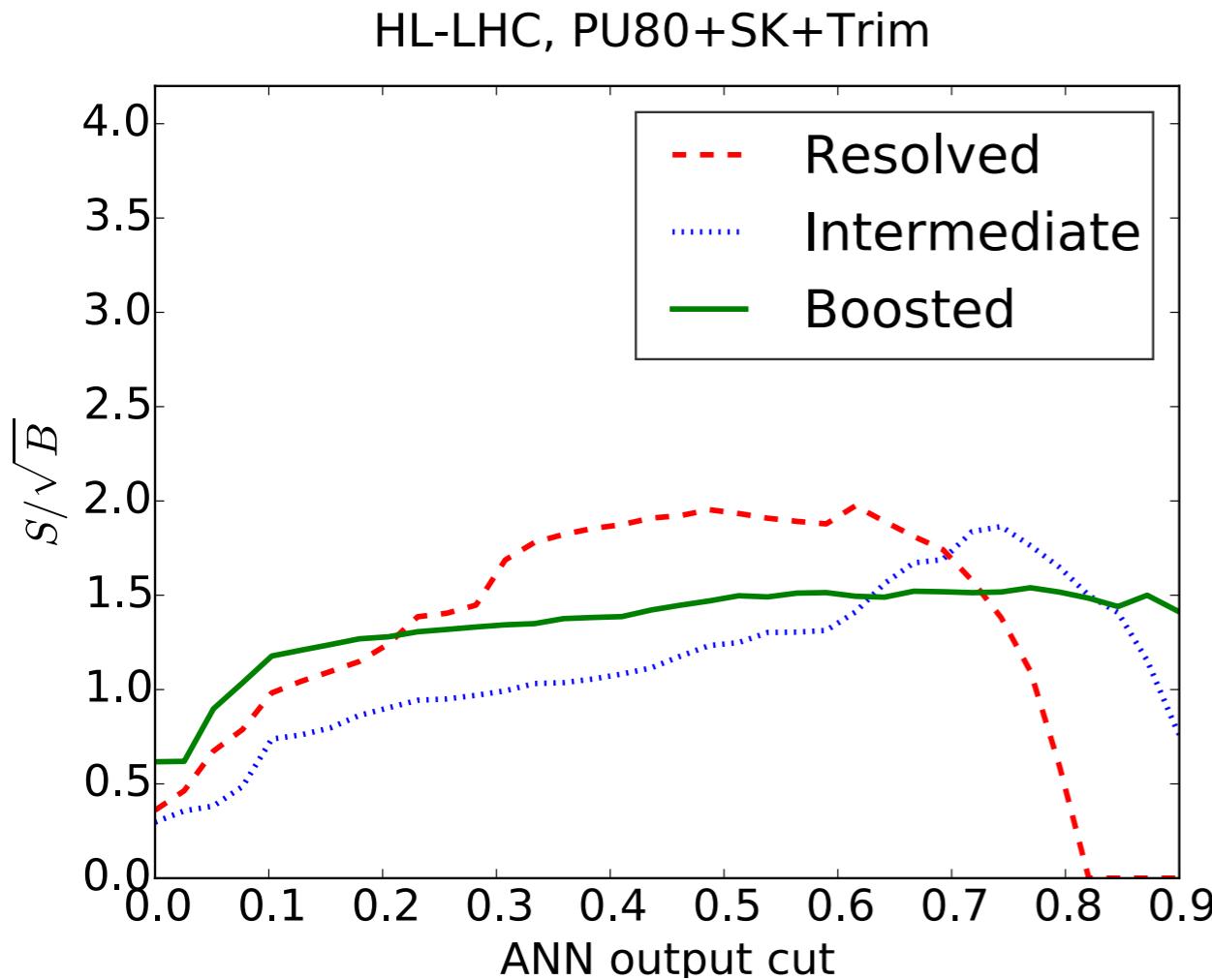


Boosted topologies most effective without pileup

Generally good balance between topologies

MULTIVARIATE ANALYSIS – RESULTS

- Vary the cut in ANN output, under the addition of pileup



Significances reshuffle - boosted topologies damaged

Boosted topologies remain less sensitive to systematics

MULTIVARIATE ANALYSIS – RESULTS

HL-LHC, PU80+SK+Trim					
Category		N_{ev} signal	N_{ev} back	S/\sqrt{B}	S/B
Boosted	$y_{\text{cut}} = 0$	410	$4.5 \cdot 10^5$	0.6	10^{-3}
	$y_{\text{cut}} = 0.8$	290	$3.7 \cdot 10^4$	1.5	0.01
Intermediate	$y_{\text{cut}} = 0$	260	$7.7 \cdot 10^5$	0.3	$3 \cdot 10^{-4}$
	$y_{\text{cut}} = 0.75$	140	$5.6 \cdot 10^3$	1.9	0.03
Resolved	$y_{\text{cut}} = 0$	1800	$2.7 \cdot 10^7$	0.4	$7 \cdot 10^{-5}$
	$y_{\text{cut}} = 0.60$	640	$1.0 \cdot 10^5$	2.0	0.01

- Combined signal significance $\sim 3.1\sigma$ after MVA (0.8σ before)

S/\sqrt{B}	no PU	PU80 + sub
pre-MVA	0.8	0.8
post-MVA	4.0	3.1

Significances for the HL-LHC
 $(\mathcal{L} = 3000\text{fb}^{-1})$

CONCLUSIONS

$HH \rightarrow (b\bar{b})(b\bar{b})$ is a **tough** process to measure, however

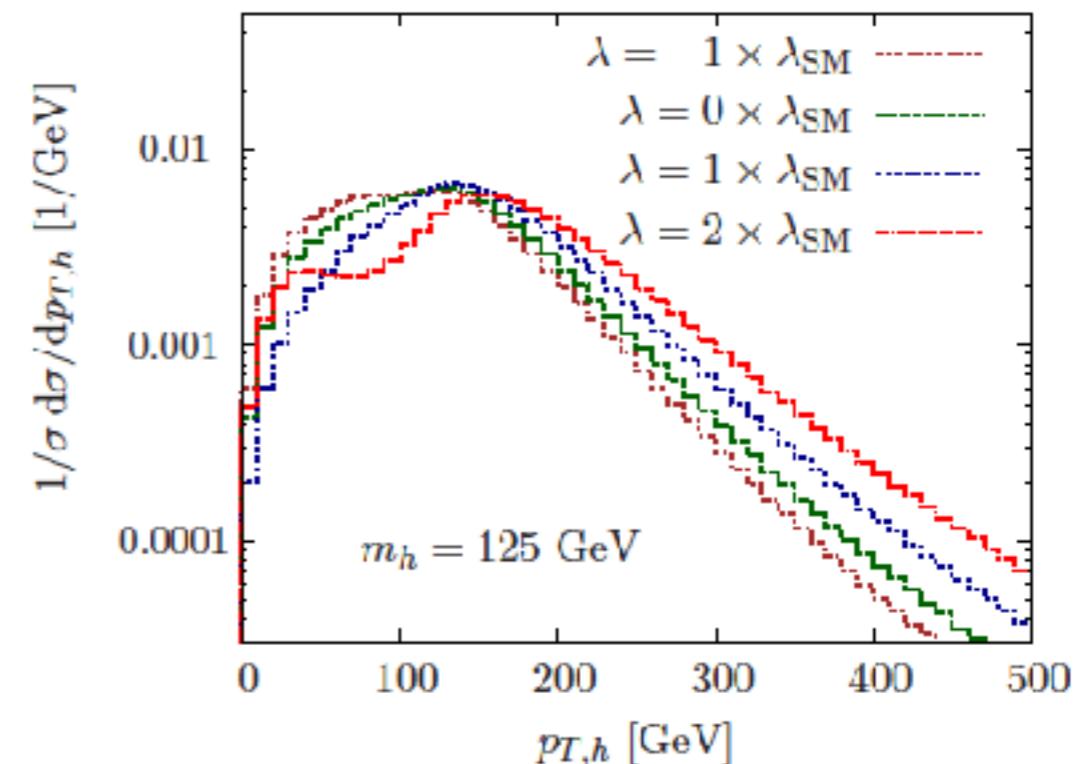
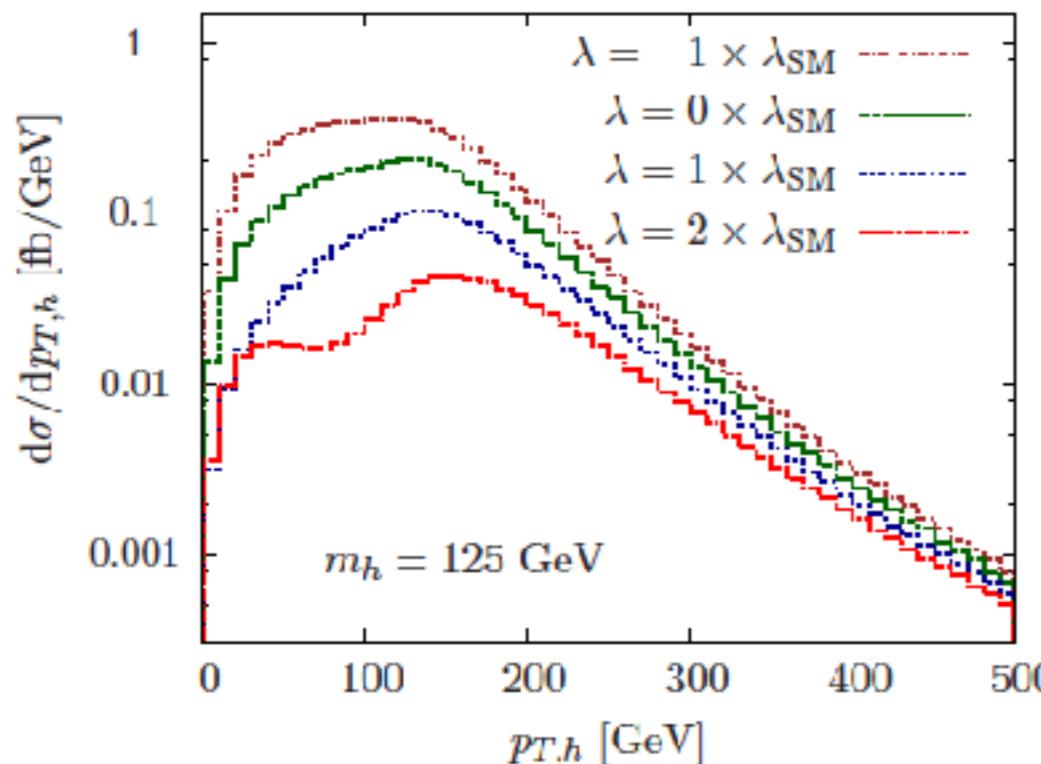
- *Multiple topologies*
- *Substructure*
- *Multivariate-analysis*

With a combined arms strategy:

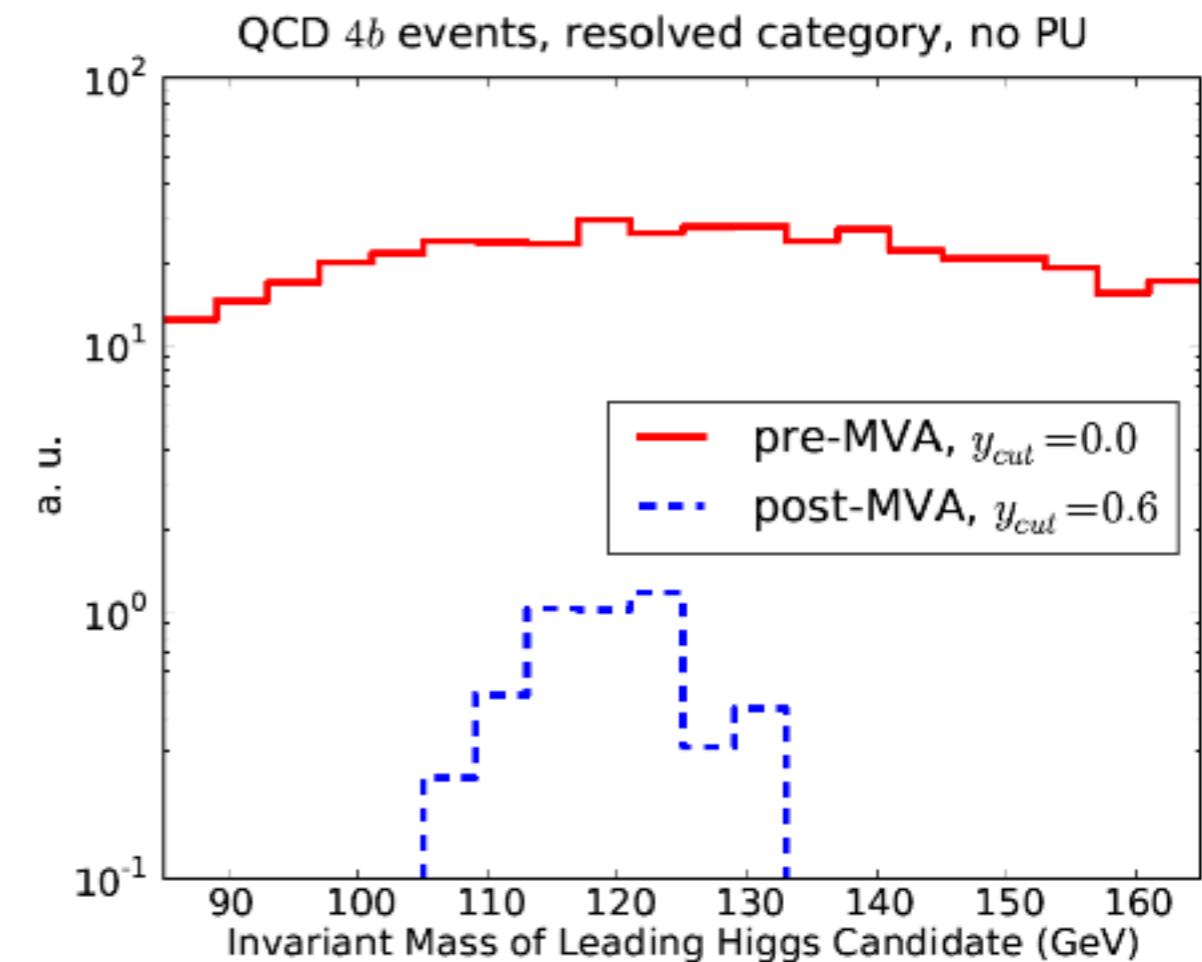
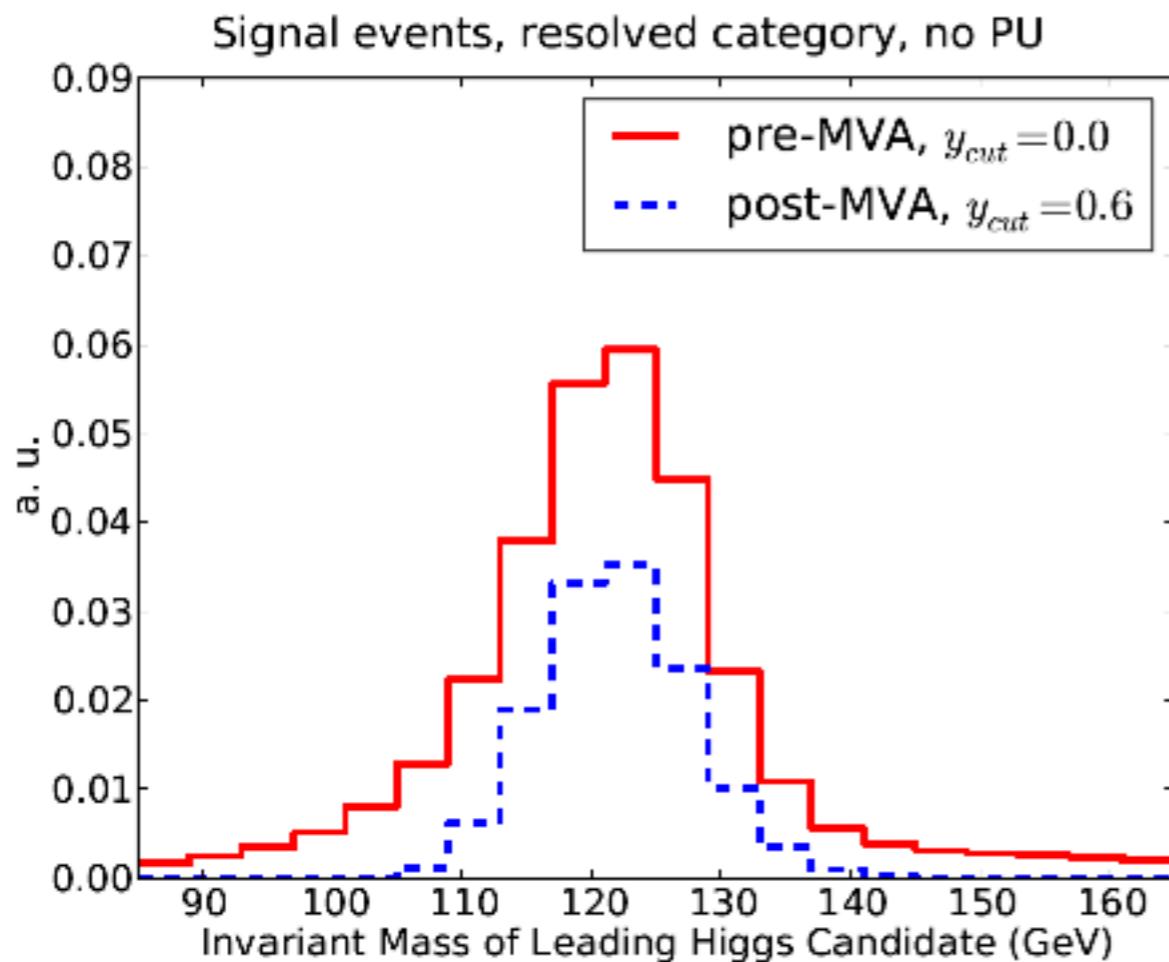
An observation of HH production in the bbbb channel is feasible at the HL-LHC

Combined $S/\sqrt{B} \sim 3.1$ after MVA

Next steps - what bounds can we obtain upon λ ?



MVA AT WORK



MULTIVARIATE ANALYSIS – TRAINING

The neural network is trained upon Monte Carlo events

An event i originates from either the **signal** $y'_i = 1$
or **background** $y'_i = 0$ processes

It carries a set of kinematic variables $\{k\}_i$
and is processed by an NN with parameters $\{\omega\}$

For each event we have the NN output: $y_i = P(y'_i = 1 | \{k\}_i, \{\omega\})$

and therefore per-event likelihood: $P(y'_i | \{k\}_i, \{\omega\}) = y_i^{y'_i} (1 - y_i)^{1-y'_i}$,

*So the optimisation consists of
minimising the Cross-Entropy*

$$E(\{\omega\}) \equiv -\log \left(\prod_i^{N_{\text{ev}}} P(y'_i | \{k\}_i, \{\omega\}) \right)$$

$$= \sum_i^{N_{\text{ev}}} [y'_i \log y_i + (1 - y'_i) \log (1 - y_i)]$$

ANALYSIS BREAKDOWN – NO PILEUP

HL-LHC, Resolved category, no PU										
			Cross-section [fb]				S/B		S/ \sqrt{B}	
	hh4b	total bkg	4b	2b2j	4j	t \bar{t}	tot	4b	tot	4b
C1a	9	$2.2 \cdot 10^8$	$6.9 \cdot 10^4$	$1.5 \cdot 10^7$	$2.0 \cdot 10^8$	$2.1 \cdot 10^5$	$4.0 \cdot 10^{-8}$	$1.3 \cdot 10^{-4}$	0.03	1.9
C1b	9	$2.2 \cdot 10^8$	$6.9 \cdot 10^4$	$1.5 \cdot 10^7$	$2.0 \cdot 10^8$	$2.1 \cdot 10^5$	$4.0 \cdot 10^{-8}$	$1.3 \cdot 10^{-4}$	0.03	1.9
C1c	2.6	$4.4 \cdot 10^7$	$1.6 \cdot 10^4$	$3.2 \cdot 10^6$	$4.1 \cdot 10^7$	$8.8 \cdot 10^4$	$6.1 \cdot 10^{-8}$	$1.6 \cdot 10^{-4}$	0.02	1.1
C2	0.5	$4.9 \cdot 10^3$	$1.7 \cdot 10^3$	$2.9 \cdot 10^3$	$2.1 \cdot 10^2$	47	$1.1 \cdot 10^{-4}$	$2.9 \cdot 10^{-4}$	0.4	0.6

HL-LHC, Intermediate category, no PU										
			Cross-section [fb]				S/B		S/ \sqrt{B}	
	hh4b	total bkg	4b	2b2j	4j	t \bar{t}	tot	4b	tot	4b
C1a	2.8	$8.4 \cdot 10^7$	$2.1 \cdot 10^4$	$5.3 \cdot 10^6$	$7.9 \cdot 10^7$	$3.3 \cdot 10^4$	$3.4 \cdot 10^{-8}$	$1.3 \cdot 10^{-4}$	0.02	1.1
C1b	2.6	$5.8 \cdot 10^7$	$1.4 \cdot 10^4$	$3.6 \cdot 10^6$	$5.5 \cdot 10^7$	$3.0 \cdot 10^4$	$4.5 \cdot 10^{-8}$	$1.9 \cdot 10^{-4}$	0.02	1.2
C1c	0.5	$3.5 \cdot 10^6$	$8.7 \cdot 10^2$	$2.1 \cdot 10^5$	$4.3 \cdot 10^7$	$8.8 \cdot 10^3$	$1.6 \cdot 10^{-7}$	$6.1 \cdot 10^{-4}$	0.02	1.0
C2	0.09	$1.8 \cdot 10^2$	56	96	22	3.1	$5.3 \cdot 10^{-4}$	$1.6 \cdot 10^{-3}$	0.4	0.6

HL-LHC, Boosted category, no PU										
			Cross-section [fb]				S/B		S/ \sqrt{B}	
	hh4b	total bkg	4b	2b2j	4j	t \bar{t}	tot	4b	tot	4b
C1a	3.9	$4.6 \cdot 10^7$	$1.1 \cdot 10^4$	$2.9 \cdot 10^6$	$4.3 \cdot 10^7$	$2.4 \cdot 10^4$	$8.2 \cdot 10^{-8}$	$3.4 \cdot 10^{-4}$	0.03	2.0
C1b	2.7	$3.7 \cdot 10^7$	$7.5 \cdot 10^3$	$2.1 \cdot 10^6$	$3.5 \cdot 10^7$	$2.2 \cdot 10^4$	$7.4 \cdot 10^{-8}$	$3.7 \cdot 10^{-4}$	0.03	1.7
C1c	1.0	$3.9 \cdot 10^6$	$8.0 \cdot 10^2$	$2.3 \cdot 10^5$	$3.7 \cdot 10^6$	$7.1 \cdot 10^3$	$2.6 \cdot 10^{-7}$	$1.3 \cdot 10^{-3}$	0.03	2.0
C2	0.16	$2.5 \cdot 10^2$	53	$1.9 \cdot 10^2$	13	1.6	$5.7 \cdot 10^{-4}$	$2.7 \cdot 10^{-3}$	0.5	1.1

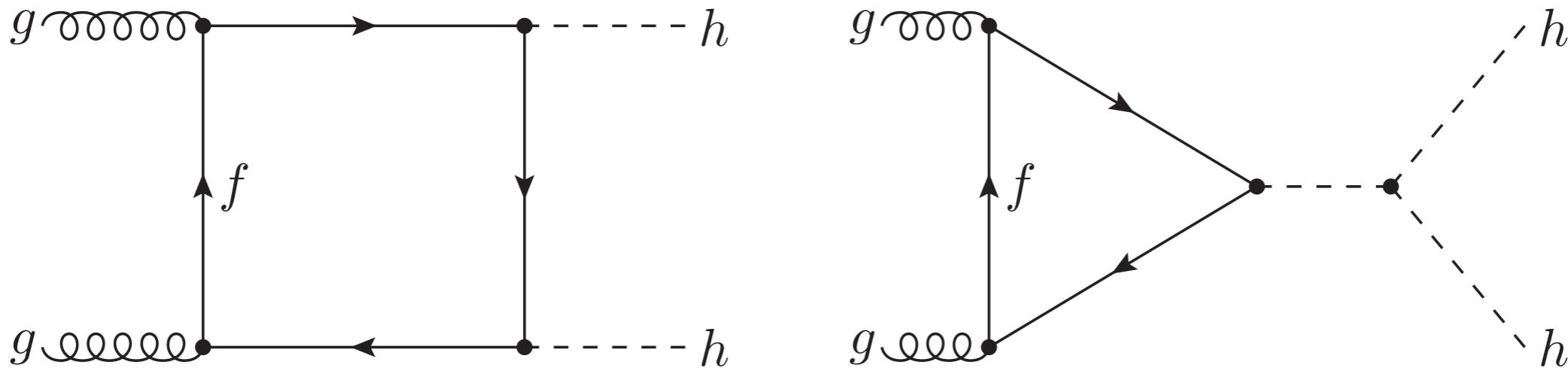
ANALYSIS BREAKDOWN - INCLUDING PILEUP

HL-LHC, Resolved category, PU+SK with $n_{PU} = 80$										
	$hh4b$	total bkg	Cross-section [fb]				S/B		S/\sqrt{B}	
			4b	2b2j	4j	$t\bar{t}$	tot	4b	tot	4b
C1a	11	$4.4 \cdot 10^8$	$1.5 \cdot 10^5$	$3.0 \cdot 10^7$	$4.1 \cdot 10^8$	$2.6 \cdot 10^5$	$2.4 \cdot 10^{-8}$	$7.2 \cdot 10^{-5}$	0.03	1.5
C1b	11	$4.4 \cdot 10^8$	$1.5 \cdot 10^5$	$3.0 \cdot 10^7$	$4.1 \cdot 10^8$	$2.6 \cdot 10^5$	$2.4 \cdot 10^{-8}$	$7.2 \cdot 10^{-5}$	0.03	1.5
C1c	3	$1.1 \cdot 10^8$	$4.2 \cdot 10^4$	$7.7 \cdot 10^6$	$9.9 \cdot 10^7$	$1.1 \cdot 10^5$	$2.8 \cdot 10^{-8}$	$7.4 \cdot 10^{-5}$	0.02	0.8
C2	0.6	$9.0 \cdot 10^3$	$3.5 \cdot 10^3$	$5.1 \cdot 10^3$	$3.1 \cdot 10^2$	50	$6.5 \cdot 10^{-5}$	$1.7 \cdot 10^{-4}$	0.4	0.5

HL-LHC, Intermediate category, PU+SK+Trim with $n_{PU} = 80$										
	$hh4b$	total bkg	Cross-section [fb]				S/B		S/\sqrt{B}	
			4b	2b2j	4j	$t\bar{t}$	tot	4b	tot	4b
C1b	2.7	$8.1 \cdot 10^7$	$2.1 \cdot 10^4$	$5.2 \cdot 10^6$	$7.6 \cdot 10^7$	$3.0 \cdot 10^4$	$3.4 \cdot 10^{-8}$	$1.3 \cdot 10^{-4}$	0.02	1.0
C1c	2.6	$6.2 \cdot 10^7$	$1.5 \cdot 10^4$	$3.9 \cdot 10^6$	$5.8 \cdot 10^7$	$2.8 \cdot 10^4$	$4.1 \cdot 10^{-8}$	$1.7 \cdot 10^{-4}$	0.02	1.1
C1d	0.5	$2.8 \cdot 10^6$	$7.9 \cdot 10^2$	$1.9 \cdot 10^5$	$2.7 \cdot 10^6$	$6.5 \cdot 10^3$	$1.8 \cdot 10^{-7}$	$6.2 \cdot 10^{-4}$	0.02	1.0
C2	0.09	$2.6 \cdot 10^2$	47	$1.8 \cdot 10^2$	30	2.2	$3.4 \cdot 10^{-4}$	$1.8 \cdot 10^{-3}$	0.3	0.7

HL-LHC, Boosted category, PU+SK+Trim with $n_{PU} = 80$										
	$hh4b$	total bkg	Cross-section [fb]				S/B		S/\sqrt{B}	
			4b	2b2j	4j	$t\bar{t}$	tot	4b	tot	4b
C1a	3.5	$4.1 \cdot 10^7$	$1.0 \cdot 10^4$	$2.7 \cdot 10^6$	$3.8 \cdot 10^7$	$2.0 \cdot 10^4$	$8.6 \cdot 10^{-8}$	$3.4 \cdot 10^{-4}$	0.03	1.9
C1b	2.5	$3.2 \cdot 10^7$	$6.8 \cdot 10^3$	$1.9 \cdot 10^6$	$3.0 \cdot 10^7$	$1.9 \cdot 10^4$	$7.8 \cdot 10^{-8}$	$3.6 \cdot 10^{-4}$	0.02	1.6
C1c	0.8	$2.2 \cdot 10^6$	$5.4 \cdot 10^2$	$1.4 \cdot 10^5$	$2.0 \cdot 10^6$	$4.8 \cdot 10^3$	$3.8 \cdot 10^{-7}$	$1.6 \cdot 10^{-3}$	0.03	2.0
C2	0.14	$1.5 \cdot 10^2$	40	86	22	1.8	$9.0 \cdot 10^{-4}$	$3.5 \cdot 10^{-3}$	0.6	1.2

DIHIGGS AT THE LHC



Clear difficulty: HH production cross-section at LHC is *tiny*

14TeV NNLO gluon fusion $\sigma_{HH} \simeq 40 \text{ fb}$ factor of $\sim 10^3$ smaller than single H

Compounded by usual H reconstruction problems

Final State	BF
bbbb	33%
bbWW	13%
bb\tau\tau	3.5%
WWWW	5.3%

Lots of cross-section decaying to challenging fully hadronic final state

Recent ATLAS bound:

$\sigma(pp \rightarrow hh \rightarrow b\bar{b}b\bar{b}) < 1.22 \text{ pb}$
(ATLAS-CONF-2016-017)

(Data is king: 3000 fb^{-1} HL-LHC)