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Boosting the standard model Higgs signal with the Template Overlap Method

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ABSTRACT: We show that the Template Overlap Method can improve the signal to background ratio of boosted $h \rightarrow b\bar{b}$ events produced in association with a leptonically decaying W. We introduce several improvements on the previous formulations of the template method. Varying three-particle template subcones increases the rejection power against the backgrounds, while the longitudinally invariant, sequential template generation ensures an efficient coverage in template phase space. We integrate b-tagging information into the template overlap framework and introduce a new template based observable, the template stretch. We further improve on the previous formulation of Template Planar Flow by including the effects of finite size subcones centered at each template momentum. Our analysis takes into account the contamination from the charm daughters of top decays in $t\bar{t}$ events, and includes nearly-realistic effects of pileup and underlying events. We show that the Template Overlap Method displays very low sensitivity to pileup, hence providing a self-contained alternative to other methods of pile up subtraction. The developments described in this work are quite general, and may apply to other searches for massive boosted objects.

KEYWORDS: Jets, Hadronic Colliders

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1 Introduction

Recent results from ATLAS [1] and CMS [2] have confirmed the discovery of a new boson of mass roughly 125 GeV, decaying to $\gamma\gamma$, ZZ^* and likely WW^* (for the Tevatron combination of the Higgs searches see ref. [3]). The results are so far consistent with the interpretation of the new particle as the Standard Model (SM) Higgs boson.

While the data agrees with the SM Higgs boson, our understanding of the new particle's properties remains incomplete. A strong test of the SM Higgs theory consists of a detailed experimental study of its sharply predicted characteristics. This includes, among others, observing all the major decay modes of the SM Higgs, as well as establishing no deviations from the predicted SM Higgs production and decay rates. Global fits to the Higgs boson

production and decay rates allow for extraction of its couplings to various other SM fields, as well as possible invisible channels (see refs. [4–8] for recent analyses). A SM Higgs boson of 125 GeV predicts a dominant decay mode to a $b\bar{b}$ pair, which calls for a direct verification. An enormous QCD background, however, makes it rather difficult to observe this channel at the LHC. The hadronic Higgs decay mode has so far only been reported by the Tevatron experiments, at 2.8 σ in the CDF/D0 combinations. Despite an impressive progress by both CMS and ATLAS, the extraction of a statistically significant measurement of $h \rightarrow b\bar{b}$ rate from the LHC data at 8 TeV remains challenging.

One way to reduce the QCD background in $h \to b\bar{b}$ is by focusing on associated Higgs production with W and Z bosons. However, the cross section of the background $pp \to Vb\bar{b}$ process is still much higher than the cross section of $pp \to Vh$. Even after a cut on the mass of the $b\bar{b}$ system around the known Higgs mass, the signal is swamped by the background. Authors of ref. [9] showed that when considering moderately boosted Higgs events, traditional jet clustering algorithms with large cone sizes $(R \sim 1)$ can be used to increase the signal to background ratio (S/B). The method of ref. [9] is based on the fact that decay products of a boosted Higgs are collimated and can be captured within a single "fat" jet. This, in principle, reduces the combinatorial background, contamination from soft and incoherent components and allows to better characterize the structure of the energy flow within the fat jet.

The last few years have seen a proliferation of new theoretical and experimental techniques to identify high- p_T jets at the LHC (see refs. [10–17] for recent reviews and references therein). Two main classes of approaches have emerged: *Filtering* [9] (see also refs. [18, 19]) and *Template Overlap Method* [20]. Filtering algorithms act on the list of jet constituents by removing the soft components based on some measure which defines the "hard" part of the jet. The remaining constituents are then reclustered into the "filtered" jet. The Template Overlap Method, discussed in detail below, does not manipulate the measured jet's list of constituents, nor does it require any special clustering algorithm. Instead the method compares the jet to a set of parton level states built according to a fixed-order distribution of signal jets called *templates*. The comparison makes use of an "overlap function" which evaluates the level of agreement between each measured jet and a set of templates.

Ref. [20] focused on building the templates according to the leading order decay modes, namely two body (N = 2) for the boosted Higgs and three body (N = 3) for the boosted top. In ref. [21], the ATLAS collaboration used the Template Overlap Method together with the the HEPTOPTAGGER [22] to search for heavy $t\bar{t}$ resonances. Authors of ref. [23] showed how to extend the Template Overlap Method beyond leading order as well as how to construct templates which describe the energy flow of say $h \rightarrow b\bar{b}g$ in an infrared (IR) safe manner. The method allows one to gain access to "partonic-like" observables which correspond to the template configurations with the maximal overlap score. The resulting information can further improve the ability to distinguish the signal from various background channels.

Furthermore, ref. [23] introduced the concept of *template jet shapes*, such as Template Planar Flow (tPf) [23–26] and template-angularity [24, 27]). Template jet shapes can further boost the background rejection power of the overlap method. The results of ref. [23]

showed that template overlap is capable of delivering background rejection factors of O(100) against the Wj's background in the idealistic ultra-boosted Higgs regime (i.e. $p_T \sim 1 \text{ TeV}$), when combined with other jet substructure observables.

In this paper, we examine the decay and radiation patterns of a boosted SM Higgs boson, with focus on a realistic p_T kinematic regime (*i.e.*, 300 - 400 GeV). We argue that a boosted Higgs search using the Template Overlap Method is viable in the future LHC run. We achieve the best signal sensitivity by combining templates in the full phase space for N = 3 and N = 2 overlaps in addition to other template based observables.

Our treatment of Template Overlap Method improves on the previous formulations in several ways. First, we define templates in terms of longitudinally boost-invariant variables. Second, and more importantly, we entirely revamp the method of template generation. In ref. [23], the minimum number of templates required to adequately describe the jet energy flow in the medium p_T range, was roughly two orders of magnitude larger than in this paper. The reason is that in ref. [23], templates were generated in the Higgs rest frame (with a MonteCarlo-like method) and then boosted to the lab frame on an event by event basis. Generating templates in the lab frame and "tiling" them according to the event kinematics leads to better coverage of phase space at lower p_T and to a great improvement in the overall performance of the analysis. Third, we introduce b-tagging into the Template Overlap framework. Information about b-jets combined with peak templates serves to improve the rejection power, defined as the signal efficiency divided by the efficiency for the background. Fourth, the optimal radius of the template subcones is not necessarily the same for every parton in a template, as low momentum subjets tend to have wider angular profiles. We allow the three-body template subcones to vary with p_T providing a more adequate description of the showering patterns within a fat jet. Varying cones improve the tagging performance of three-body overlap as well as most of the other template-derived observables. Fifth, we introduce a new variable, Template Stretch, which exploits the difference in plain distance of the two leading b-tagged subjets relative to the signal expectations. Template stretch is correlated with the jet mass, but with an important advantage of robustness against pileup. Finally, we include the template subcones in the formulation of tPf. Inclusion of subcones serves to simulate the effects of showering on the numerical value of tPf.

Our analysis includes nearly-realistic effects of pileup and underlying event (UE). In high luminosity environments, the large jet cone radius allows for severe effects of pileup on the spectrum of both jet and substructure observables. We show that the template jet shapes give us an additional handle on pileup, as they are based on best matched templates and not jet constituents. The "spikiness" of the jet energy distribution naturally avoids the complication of soft un-correlated backgrounds. The Template Overlap Method is thus less susceptible to pileup compared to other kinematic observables such as jet invariant mass and p_T . This feature invites us to re-consider several jet-shape observables in terms of the "partonic" distribution of the peak templates. For instance, jet Planar Flow (Pf) is known to exhibit high susceptibility to pileup [28, 29], however it is a useful background discriminant when the hard and coherent part of the massive jet is considered [24, 25, 30]. We thus introduce a pileup-insensitive alternative to Planar Flow constructed from the template states alone. We use parton-shower simulations to illustrate the non-susceptibility of the Template Overlap Method to a high-pileup environment. Our results agree with the 7 TeV ATLAS data analysis in ref. [21], which showed that the overlap method is indeed fairly robust to presence of moderate pileup contamination.

In section 2, we give an overview of the Template Overlap Method and introduce several template based observables sensitive to the QCD radiation patterns. Section 3 describes our Monte Carlo (MC) data generation, and shows the results for boosted Higgs searches with a mass of 125 GeV at $\sqrt{s} = 8$ TeV and $\sqrt{s} = 13$ TeV. Section 3 also contains a detailed discussion of pileup effects. We give a detailed review of template Planar Flow in appendix A, while appendix B describes the new method of template states generation.

2 Template Overlap Method

Template Overlap Method is based on the quantitative comparison between the energy flow inside physical jets and the energy carried by partons modeled after the boosted signal events (templates). We define libraries of templates as sets of N four-momenta $|f\rangle = |p_1, p_2, \dots, p_N\rangle$ representing the decay products of a SM Higgs boson at a fixed momenta P and Higgs mass m_h :

$$\sum_{a=1}^{N} p_a = P, \qquad P^2 = m_h^2. \tag{2.1}$$

We require that the N quanta of energy be captured within an anti- k_T jet of varying size, scaled according to the p_T of the Higgs (or according to the p_T of the associated vector boson). The number of partons in the templates is calculated at fixed order in perturbation theory,¹ but is not limited to the leading order configuration, and "next-to-leading-order" templates with more than the minimum number of partons are possible. Below we focus on combining the information from templates in the full phase space for N = 3 and N = 2partonic configurations, and show that they provide some additional rejection power against the dominant $Wb\bar{b}$ and $t\bar{t}$ backgrounds.

Next, a functional measure quantifies the agreement in energy flow between a given Higgs decay hypothesis (a template) f and an observed jet j. A scan over a large set of templates that cover the N-body phase space of a Higgs decay results in f[j], the template which maximizes the functional measure. Our primary jet substructure observable is the N-body overlap Ov_N , the value of the functional measure for the best matched template.

For each jet candidate, we define the (maximum) overlap as

$$Ov_N(j, f[j]) = \max_{\{f\}} \left[\exp\left[-\sum_{a=1}^N \frac{1}{\sigma_a^2} \left(\epsilon \, p_{T,a} - \sum_{i \in j} p_{T,i} \, F(\hat{n}_i, \hat{n}_a) \right)^2 \right] \right], \quad (2.2)$$

where $\{f\}$ collectively denotes a template library for the given jet p_T , $p_{T,a}$ is the transverse momentum of the a^{th} template parton and $p_{T,i}$ is the transverse momentum of the i^{th} jet

¹In principle, one can re-sum the soft radiation from each of the partons. Here we simply stick to a fixed order perturbation theory description.

constituent (or calorimeter tower, topocluster, etc.). The first sum is over the N partons in the template and the sum inside the parentheses is over jet constituents. The kernel functions $F(\hat{n}, \hat{n}_a)$ restrict the angular sums to (nonintersecting) regions surrounding each of the template momenta. We refer to the template state which maximizes the functional measure as the "peak template" f[j] of the jet j. The parameter ϵ allows us to correct for the energy not captured by the template overlap. In the case of N = 3, we use $\epsilon = 0.8$. For more details see section 2.2.

The definition of Ov_N in eq. (2.2) differs from the previous definitions of overlap in refs. [20, 23]. We find that the formulation of Ov_N in terms of longitudinally boost-invariant quantities is more appropriate for a detailed hadron collider study. Notice, however, that since we will be focusing on Higgs events with high transverse momentum, the differences between the results obtained with previous definitions of Ov_N and eq. (2.2) should not be significant.

In this analysis, we take the kernel function to be a normalized step function that is nonzero only in non-overlapping angular regions around the directions of the template momenta p_a :

$$F(\hat{n}_i, \hat{n}_a) = \begin{cases} 1 & \text{if } \Delta R < r_a \\ 0 & \text{otherwise} \end{cases},$$
(2.3)

where ΔR is the plain distance between the template parton and a jet constituent in the (η, ϕ) plane. The parameters r_a determine the angular scale of the template subjet. Together with the energy resolutions σ_a , these are the only tunable parameters of the model.

The kernel definition of eq. (2.3) assumes that the angular regions around the template momenta are non-intersecting, meaning that any pair of template momenta i, j satisfies the condition

$$\Delta R_{ij} > r_i + r_j,$$

where $r_{i,j}$ are the template subcones. Eq. 2 serves to ensure that a single particle in an event can not be assigned to more than one template momentum in the overlap calculation. If the condition for non-intersecting templates is not satisfied, the template is discarded as non-valid.

A few ideas for possible strategies to determine the values of r_a and σ_a are listed below:

- Choose the single best parameter according to some optimization criterion (e.g., optimize the tagging efficiency and background rejection), and use the same values for all the partons within a set of templates.
- Choose the parameters separately for each template, e.g. using a p_T -dependent scale for template matching.

Based on a combination of these two criteria, we fix σ_a (for the a^{th} parton) by that parton's transverse momentum,

$$\sigma_a = p_{T,a}/3. \tag{2.4}$$

We use a subcone of radius $r_2 = 0.3$ for the two-body template analysis, while the threebody subcone is dynamically determined on a template-by-template basis. Section 2.2 contains a detailed discussion on the optimal scaling of subcone radii. We should nevertheless emphasize that the overall performance of the method can be maximized for a wide range of r by rescaling other parameters.

Next, we generate libraries of $O(10^4)$ 2-body templates and $O(10^6)$ 3-body templates in steps of Higgs p_T of 30 GeV starting from 315 GeV. For each event, we achieve the best signal sensitivity by dynamically selecting a set of templates based on the transverse momentum of the W boson,

$$p_T^W \in \left[p_{T\min}^{\min}, \, p_{T\max}^{\min} \right), \tag{2.5}$$

where $p_{T\min}^{\text{bin}}$, $p_{T\max}^{\text{bin}}$ are the limits of the bin corresponding to the template set with $p_T^{\text{temp}} = (p_{T\max}^{\text{bin}} - p_{T\min}^{\text{bin}})/2$. For instance, an event with $p_T^W = 320 \text{ GeV}$ would be analyzed by a template set with $p_T = 315 \text{ GeV}$ etc. We give more details on template properties and generation in the appendix.

Replacing the criteria for the template set selection from jet p_T to p_T^W has a enormous advantage when considering effects of pileup, however it is not the only choice. As an alternative, one could analyze each jet with all template sets, but at a huge expense in computation time.

We use the TEMPLATETAGGER [31] numerical package for the template matching analysis.² The package is a C++ code which provides basic implementation of the Template Overlap Method for jet substructure, as well as several other jet analysis tools.

2.1 Other peak template observables

Template overlap provides a mapping of final states j to partonic configurations f[j] at any given order in perturbation theory. Once the best matched template f[j] is found, it can be used to characterize the energy flow of the state, giving information on the likelihood that the event is signal or background. The scope of template overlap does not stop with Ov_2 and Ov_3 . Peak templates contain additional information about correlations within the fat jet, some of which we explore in the following sections.

2.1.1 Angular correlations

Of particular value are angular correlations between template momenta which can otherwise be concealed in the numerical values of the peak overlap. For instance, the angular distribution of jet radiation can be measured with the variable $\bar{\theta}$ [23], defined as

$$\overline{\theta} = \sum_{i} \sin \Delta R_{iJ}, \qquad (2.6)$$

where ΔR_{iJ} is the distance in the (η, ϕ) plane between the i^{th} template partonic momentum and the jet axis. When measured using three-body templates, the $\bar{\theta}$ variable exploits the fact that signal events tend to have smaller emission angles. Notice that for highly boosted jets, the 2-body version of $\bar{\theta}$ simply reduces to the angle between the two partons [24].

²Publicly available at: tom.hepforge.org.

2.1.2 Template Planar Flow

Another useful background discriminant is Planar Flow [24, 25] (see ref. [26] for a recent study of the Pf distribution of QCD massive boosted jets). Previous definitions of Pf were defined in terms of the two momenta in the plane perpendicular to the jet axis. Here, we choose to define the Pf variable in terms of longitudinally boost-invariant quantities. We begin by defining the two-momenta p_i^k as

$$p_i^1 \equiv p_i^T \eta_i, \qquad p_i^2 \equiv p_i^T \phi_i, \tag{2.7}$$

with η_i , ϕ_i measured relative to the jet axis.

Next, we define the "jet inertia tensor" I^{kl} as:

$$I^{kl} \equiv \frac{1}{m_J} \sum_{i}^{N} \frac{p_i^k p_i^l}{p_i^T},\tag{2.8}$$

where m_J is the jet mass, *i* runs over the jet constituents and $p_i^{k,l}$ are defined in eq. (2.7). The Pf jet shape is then given by

$$Pf \equiv \frac{4 \det(I)}{\operatorname{tr}(I)^2} \,. \tag{2.9}$$

Notice that because of the fact that the trace of (I^{kl}) is proportional to the jet mass [32], Planar Flow is only well defined for massive jets [29]. Planar flow is particularly helpful in distinguishing energy flow distributions which lie on a line (Pf $\rightarrow 0$) from uniformly distributed energy flow (Pf $\rightarrow 1$). For instance, Planar Flow of a boosted Higgs will tend to be smaller than that of a massive QCD jet that in turn will be smaller than that of a boosted top or gluino [33].

Planar Flow of a *jet* is useful when considering only the hard and coherent part of the jet. Because of high sensitivity of jet Planar Flow to pileup and UE, here we consider Template Planar Flow (tPf) [23], as an alternative. We define tPf using peak template momenta as well as the template subcones to include physical effects of energy smearing. To compute tPf, we use the definition of eq. (2.9) with two major differences. First, in the definition of tPf, the index *i* runs over the three template momenta of a three-body template configuration. Second, the inertia tensor takes the form

$$I^{kl} \to I_s^{kl} = I^{kl} + \frac{1}{m_J} \sum_{i=1}^3 p_i^T r_i \,\delta^{kl}, \qquad (2.10)$$

where r_i is the template subcone radius and I^{kl} is defined in eq. (2.8). Note that other definitions of I_s^{kl} are possible. Appendix A gives a more detailed discussion.

The advantage of tPf is that it is constructed purely out of peak template states and thus less susceptible to pileup. We demonstrate this point using a Monte Carlo simulation in section 3.2. In addition, the non zero mass of the peak template states guarantees infrared safety of tPf, whereas jet Pf is infra-red safe only after a cut on the jet invariant mass is applied.

2.1.3 Template b identification

A simple but useful way to reject backgrounds and accept signal events is to incorporate the information related to *b*-tagging into the template overlap framework. The identification of *b*-tagged jets relies on information beyond what is provided by the calorimeters (say from the presence of a displaced vertex and/or a hard lepton). This information (e.g. direction of the *b*-tagged jets in η and ϕ) is fairly uncorrelated with the information about the direction of the peak template partons. We integrate the *b*-tagging information into the template overlap framework by assigning a *b*-quark tag, $t_b^{(f)}$ to each peak template f[j]. A two body template parton is assigned a *b*-tag if an anti- k_T (r = 0.4) *b*-jet lies within a template cone of radius r_2 around the template parton axis. For simplicity, we take a jet to be *b*-tagged if it has $p_T > p_T^{\text{tag}}$ (default: $p_T^{\text{tag}} = 20 \text{ GeV}$) and contains a *b* or \bar{b} quark.

Information about template b-tags can be of particular use in discriminating the large $t\bar{t}$ background. For instance, consider a typical doubly-b-tagged jet coming from a $t\bar{t}$ event. A fragment of another light (or c) jet is likely to fall in the cone of $R \sim 1$. If only the criterion of a doubly b-tagged jet is used, there is no guarantee that the peak two body template will select the two b subjets, making the $t\bar{t}$ event more likely to pass the kinematic constraints of the template states. On the contrary, if we require that the b-tagged jets coincide with the template momenta, we discriminate against jets in which only one b-jet is tagged by the template. Notice that the effect of b-tagging on the templates should not have a large effect on the signal.

2.1.4 Template Stretch

In order to increase the signal efficiency of the Tempate Overlap Method we chose a working point where the energy resolution of each of the template partons is rather loose (i.e. $\sigma_a = p_{T,a}/3$). This implies that even after both Ov_2 and Ov_3 cuts the mass distribution for the background events is still broad and certainly more spread than that of the signal. This, as well as the fact that for template b-tagging we require a large r = 0.4 anti- k_T jet parameter (motivated by the current experimental defaults), implies that the angular distance between the partonic, "b" candidates with a high Ov_2 score would still have some smearing with respect to the actual distance between the two anti- k_T b-tagged jets. To capture this effect, we define a new observable, *Template Stretch*, as

$$S_{b\bar{b}}^{(t)} = \frac{\Delta R_{b\bar{b}}}{\Delta R_t}.$$
(2.11)

where ΔR_t is the distance between the peak two-body template momenta and $\Delta R_{b\bar{b}}$ is the distance between the two *b*-tagged subjets. We expect that the background events will have a broader distribution of $S_{b\bar{b}}^{(t)}$ compared to the signal events. The functionality of the template stretch is correlated with the mass of the jet, but with an important advantage. The mass of a fat jet is subject to a large jet cone radius $R \sim 1$, making it highly susceptible to effects of pileup and underlying event. Since $S_{b\bar{b}}^{(t)}$ is constructed out of subjets with r = 0.4 and template states, it is bound to be less sensitive to pileup. We will illustrate this point further in section 3.2 (See figure 13).

2.2 Varying subcone templates and showering correction

Fixed template cones are limited by the fact that different p_T subjets yield a different energy profile in η , ϕ . This fact is important for three body template analysis, where we expect the p_T of the three *peak* template partons to be non-uniform. For instance, one would find that a template subcone of radius r = 0.05 is adequate to capture the radiation pattern of a 500 GeV quark. Yet, the same subcone would completely fail to adequately describe the radiation pattern of a quark with energy of 100 GeV, resulting in a poor overlap score. We thus introduce the concept of scaled three body subcones into the template overlap framework. Varying template subcones allow us to correct for energy deposition outside the template subcone radii. This in turn leads to an an improved template-level energy resolution while keeping systematic uncertanties well under control.³ Jets become narrower as p_T increases, meaning that a smaller jet area is needed to collect some fixed fraction of the jet energy at higher transverse momentum. The corresponding distribution of jet areas (at a fixed energy fraction) is generically non calculable but is measured by experiments, and is commonly denoted as the "jet shape" variable. Refs. [34] and [35] present ATLAS and CMS studies of the jet shape variable respectively.

We used the ATLAS jet shape study in ref. [34] to establish a scaling rule for the template subcones. The differential jet shape drops rapidly as r increases: at low p_T , more than 80% of the transverse momentum is contained within a cone of radius r = 0.3 around the jet direction. This fraction increases up to 95% at very high p_T . We fit the numerical values of the integrated jet shape in different p_T regions to obtain the minimum cone radius required to capture 80% of the jet transverse momentum. In the overlap analysis, we correct for the 80% efficiency by scaling the template p_T accordingly. Figure 1 shows the result of our dynamical scaling. The points represent the minimum radius necessary to capture 80% of jet's energy as a function of jet p_T . The resulting curve gives a shape to the optimal scaling rule for template subcones. The error bars on the data points are small enough that they can be omitted for the purpose of our analysis. To obtain the subcone values in the $p_T < 30$ GeV region, we extrapolate the data.

An overall shift in the optimal jet-shape curve remains a free parameter. To calibrate it we choose a benchmark point of $r_3(100 \text{ GeV})$. We demonstrate in the following section that other calibrations are possible and may perform in a similar manner, within a reasonable range. The dotted curve shows that a shift of $\delta r = 0.063$ units provides an excellent fit. Notice that the naive scaling

$$r_3(p_T^t) = r_3(100 \text{ GeV}) \frac{100 \text{ GeV}}{p_T^t},$$
 (2.12)

is in excellent agreement in the $p_T > 60$ GeV region, while the discrepancy with data becomes large at lower jet p_T .

³At low p_T , the leakage of QCD radiation outside the template subcones can be especially large when using fixed subcones. To account for this, one has to include energy correction factors into the p_T of the templates, which are largely affected by systematic uncertainties.



Figure 1. Jet shape results on optimal template subcone scaling. The blue, solid curve shows the minimal anti- k_T radius necessary to capture 80% of jet energy, obtained from ATLAS data. The grey, dashed curve is the ATLAS result shifted down by $\delta r = 0.063$, in order to match the optimized value of $r_3(100 \text{ GeV}) = 0.12$ (large, red dot). The gray shaded area shows the region where no published data exists. The short extrapolation into the shaded region was performed while imposing an upper limit on $r_3 < 0.3$. The orange, dotted line is the naive scaling of eq. (2.12).

Note that the extrapolation procedure in figure 1, while somewhat arbitrary, also does not affect the results of our analysis. We impose the condition that

$$0.05 \le r_3 \le 0.3 \tag{2.13}$$

for all template momenta. The lower limit on r_3 takes into account the detector granularity, while the upper limit provides consistency with the two body template analysis. In addition, we find that the shape of the optimal curve crosses the maximum value of $r_3 = 0.3$ a bit below $p_T = 30$ GeV, keeping the method largely insensitive to the extrapolation procedure. In the ultra high energy regime, the extrapolation also becomes irrelevant, as the template cone size becomes limited by the detector resolution.

Figure 2 shows an example of the effect of varying subcones on peak templates compared to fixed ones. The blue, dotted circles represent the peak two body template with radius $r_2 = 0.3$. The red, dashed circles are peak three body templates with a fixed $r_3 = 0.12$ (left panel) and varying r_3 (right panel). Notice that a higher percentage of the lowest p_T subject is "encompassed" by the varying cone resulting in an overall increase in the Ov_3 score.

Varying subcones improve the performance of templates on the distribution level as well. Figure 3 shows a comparative example for Ov_3 . The panel on the left was obtained using a fixed $r_3 = 0.12$ while r_3 was allowed to vary according to the scaling rule of figure 1 in the right panel. Notice that the varying subcones result in background overlap distributions which are significantly more peaked in the region of low overlap values. On the other hand the signal remains mildly affected, resulting in improved performance.



Figure 2. Template analysis of a boosted Higgs jet clustered with a radius of R = 1.2. Left panel shows a Higgs jet analyzed with fixed three body subcones of radius $r_3 = 0.12$. The right panel corresponds to a repeated the analysis using the scaled subcones shown in figure 1. Grey squares represent the jet constituents, the p_T of which is proportional to the size of the square. The solid circles are positions of *b*-quarks in the hard process.



Figure 3. Template observable distributions for doubly-b-tagged events with $Ov_2 > 0.8$ and btagged 2-body templates ($r_2 = 0.3$) for $p_T^W > 300$ GeV and no mass cut. Jets are reconstructed using FASTJET, and the anti- k_T algorithm with a varying cone size R (here denoted "ankt". See more details on R scaling below). The left panel represents distributions with $r_3 = 0.12$. The right panel represents the same distributions using the scaled subcones of figure 1.c In both panels, the fat jet was clustered with the anti- k_T algorithm and R = 1.2.

2.3 Stability of the Template Overlap Method

How sensitive is the Template Overlap Method rejection power to the choice of subcone radius $r_{\text{shape}} \equiv r_3(100 \, GeV)$ and the overall normalization of the working curve of figure 1? Significantly modifying the overall scale for subcone radius r_{shape} does alter the distribution of our kinematic variables; in particular, reducing the size of r_{shape} tends to shift Ov_2 and Ov_3 to lower values and vice versa. The sensitivity of the template method to the specific choice of r_{shape} calls for a dedicated experimental study on a control sample to fix the corresponding value. One can use either boosted top analysis in the signal area



Figure 4. Sensitivity of Wbb rejection power to the choice of radius $r_{\text{shape}} = r_3(100 \text{ GeV})$. The blue line shows rejection power with a rigid $Ov_3 > 0.6$ cut. The green line is the rejection power with a cut on Ov_3 varied with r_{shape} , while keeping a fixed efficiency. The rejection power is relative to cross sections with no Basic Cuts (see next section).

(in fact the Ov_3 distribution of boosted tops and their backgrounds was already studied experimentally by ATLAS in ref. [21]) or look at hadronic W inside a boosted top jet as a way to experimentally analyze the above dependence. The sensitivity to $r_{\rm shape}$ is of course not unique to our proposal, as even if fixed cones are used, all jet substructure methods will depend on the choice of the corresponding subcone parameters. Varying the subcone size to keep the enclosed energy fraction fixed is in fact a more covariant way to proceed with the substructure analysis. We further wish to point out that it is in general quite possible to choose values of cuts on Ov_2 and Ov_3 such that the overall signal and background efficiencies are essentially the same for different subcone radii. To illustrate this point, in figure 4 we show curves of rejection power for several choices of the subcone parameter r_{shape} . The meaning of r_{shape} in figure 4 is defined as the choice of $r_3(100 \text{ GeV})$, while the actual value of r_3 is allowed to vary according to the scaling rule in figure 1. The blue squares represent rejection power against the Wbb background as a function of the threebody subcone radius $r_3(100 GeV)$, and a fixed $Ov_3 > 0.6$ cut, while the Ov_3 cut is allowed to scale with $r_{\rm shape}$ to provide a five percent efficiency for each $r_{\rm shape}$. Signal and background efficiency of a fixed Ov_3 cut shift with the choice of r_{shape} , but disproportionately. Rigid Ov_3 cuts thus show high sensitivity to the choice of r_{shape} as shown by the blue squares in figure 4. Alternatively, fixing signal efficiency also fixes the background efficiency for a wide range of r_{shape} , thus preserving the rejection power over a wide range of r_{shape} (i.e. $r_{\rm shape} < 0.12$).

3 Data simulation and analysis

In this section, we investigate the tagging efficiencies for Higgs jets and the mistag rates for QCD backgrounds using template overlap. We consider events in which a boosted $h \rightarrow b\bar{b}$ jet is produced in association with a leptonically decaying W boson (only first two generations of leptons). The most dominant backgrounds for this process come from $t\bar{t}$ and $Wb\bar{b}$, while other channels do not significantly contribute after b-tagging requirement.

The scope of our analysis includes data at both $\sqrt{s} = 8$ TeV and $\sqrt{s} = 13$ TeV with and without pileup. We use MG/ME v5.1.33 [36] interfaced to PYTHIA v6 [37] (with MLM matching [38]) as well as SHERPA v1.4.0 [39] (with CKKW matching [40]) for data generation and the CTEQ6L1 [41] parton distribution functions. We perform jet clustering with a FASTJET [42] implementation of the anti- k_T jet clustering algorithm. To simplify the notation from here on we solely refer to PYTHIA and SHERPA results and leave the fact that all our data samples are matched implicit. We simulate Wh, $Wb\bar{b}$ and $t\bar{t}$ samples using PYTHIA, while SHERPA simulations of Wh and $Wb\bar{b}$ data serve to illustrate the effects of various showering algorithms by comparison. We only include the Wh and $Wb\bar{b}$ data in the comparison, as the $t\bar{t}$ distributions are characterized by hard scales and therefore less sensitive to detail of the showering. We scale the fat anti- k_T jet radius according to the Wmomenta

$$R = \max\left(1.4 \,\frac{200 \,\,\text{GeV}}{p_T^W}, 0.8\right),\tag{3.1}$$

where p_T^W is the transverse momentum of $p^l + p_T'$. Note that the momentum of the W is highly correlated with the momentum of the Higgs as they recoil against each other. However, scaling the cone according to p_T^W has an advantage in that it is not susceptible to pileup.

Continuing, we limit the value of R from below to be higher than 0.8 as to be able to accommodate two 0.4 anti- k_T jets used for b-tagging. The scaled fat-jet cone fulfills three tasks; it is designed to capture the $b\bar{b}$ at a fixed efficiency rate of ~ 80%; it reduces the amount of contamination from soft radiation and pileup for events with high p_T Higgs jets; and it lowers the overall $t\bar{t}$ background. Note that the scaling rule of eq. (3.1) has only a minor effect on the correlation between the Higgs fat jet momenta and that of the W as is shown in figure 5.

Next, we normalize our data to next-to-leading-order (NLO) cross sections obtained from MCFM 6.3 [43]. The cross sections assume a fixed renormalization/factorization scale of $\mu = p_T^{\min} = 300$ GeV at 8 TeV and $\mu = p_T^{\min} = 350$ GeV at 13 TeV and CTEQ6.6M [44] parton distribution functions. For each event, we find the jet with the highest transverse momentum j and impose the following *Basic Cuts*:

$$p_T^j > p_T^{\min}, \qquad p_T^W > p_T^{\min}, \\ \eta_j, \eta_l < 2.5, \qquad p_T > 40 \text{ GeV}, \\ N_b = 2, \qquad \Delta R_{bb} \ge 0.4, \\ N_k(p_T > 20 \text{ GeV}) < 2, \qquad N_l(p_T > 20 \text{ GeV}) = 1, \qquad (3.2)$$



Figure 5. The left panel shows the correlation between the Higgs-jet p_T and the p_T of the associated W for a cone of fixed size R = 1.4. The right panel shows the same correlation with the jet cone rescaled according to p_T^W as in eq. (3.1).

fb	$t\bar{t}$	$W b \overline{b}$	Wh	S/B
$\sigma(\sqrt{s} = 8 \text{ TeV}, p_T^W > 300 \text{ GeV})$	565.0	56.0	1.6	
$\sigma(\sqrt{s} = 8 \text{ TeV}, \text{ Basic Cuts})$	2.0	2.5	0.2	0.05
$\sigma(\sqrt{s} = 13 \text{ TeV}, p_T^W > 350 \text{ GeV})$	956.0	47.0	1.2	
$\sigma(\sqrt{s} = 13 \text{ TeV}, \text{ Basic Cuts})$	3.0	1.7	0.3	0.06

Table 1. NLO signal and background cross sections at $\sqrt{s} = 8$ TeV and $\sqrt{s} = 13$ TeV. The listed numbers assume a leptonically decaying W^{\pm} with first 2 generations of leptons included. Basic Cuts include eq. (3.2) as well as the *b*-tagging efficiencies.

where $N_{k,l}$ is the number of jets (anti- k_T , r = 0.4) and leptons outside the highest p_T fat jet (of radius R, determined according to eq. (3.1)) and N_b is the required number of b-tagged (anti- k_T r = 0.4) subjets. For a fat anti- k_T jet j (of radius R, determined according to eq. (3.1)), and an anti- k_T , r = 0.4 jet k, a jet is considered to be outside the fat jet if the plain distance $\Delta R(j,k) > R + r$. Similarly, a lepton l is considered outside if $\Delta R(j,l) > R$. Table 1 summarizes the cross section results with and without Basic Cuts.

We consider $p_T^{\min} = 300,350$ GeV respectively at 8 TeV and 13 TeV. For *b*-tagging we assume an efficiency of 75% and fake rate of 1% for light jets [45]. The current studies suggest a charm fake rate of 18% [45], which is likely a conservative estimate. Charms are extremely important when considering boosted Higgs decays, as the largest part of the $t\bar{t}$ background comes from events in which one top decays leptonically, while the hadronic W from the other top decays to a charm. We emphasize that omitting the charms as a source of background (as is done in some of the boosted Higgs analyses) will result in an improved performance for our tagger. Yet, at present, it is not clear whether this is possible, and the burden of proof is thus placed on the experimental collaborations.

We analyze the cases with and without pileup separately in order to illustrate the sensitivity of the Template Overlap Method to a pileup environment. This allows us



Figure 6. Invariant-mass peak searches with templates. The left panel shows the mass distributions with no cuts on template observables. The right shows the same distributions when overlap cuts are applied to the data: $Ov_2 > 0.9$ and $Ov_3 > 0.8$.

to determine the range of background rejection power as a function of the efficiency of pileup subtraction. In addition, it also allows for a comparative study of the various jet substructure observables in a pileup environment.

3.1 Higgs tagging with Template Overlap - no pileup

We proceed to discuss the ability of the Template Overlap Method to discriminate between different sources of coherent QCD radiation.

In terms of pileup filtering, analysis without pileup is equivalent to stating that the efficiency of pileup subtraction is 100%. In this section we present only the results on jets with $p_T > 300$ GeV, simulated at $\sqrt{s} = 8$ TeV, while we postpone the discussion of the future 13 TeV LHC run until section 3.1.2. An important feature of the Template Overlap Method is that it is designed to identify a particular kinematic jet substructure configuration, including the jet p_T and mass. High peak overlap score implies that the kinematics of a fat jet matches the kinematics of the peak template state. In figure 6 we plot the jet mass distribution without (left panel) and with (right panel) template overlap cuts of $Ov_2 > 0.9$ and $Ov_3 > 0.8$, after the Basic Cuts of eq. (3.2) have been applied. It is evident from figure 6 that sizable chunk of the background is removed as a result of the overlap cuts though the resolution of the fat jet Higgs mass is only moderately improved.

The mass resolution of the peak templates depends largely on the chosen parameters of the method, namely the template cone radii r_a and their energy resolution σ_a which we choose rather loosely as to keep the signal efficiency at a reasonable level. Recently, authors of ref. [21] presented a jet substructure analysis of boosted $t\bar{t}$ pairs at ATLAS. Their results showed that even with a high peak overlap cut, an additional mass window improved the background rejection power by a factor of two. Our result in figure 6 agrees with the ATLAS result. It appears that even after the overlap cuts, a mass window of (say) 110 GeV $< m_j < 130$ GeV would improve the background rejection power (this would, however, require an additional procedure of pileup removal). In the following sections we remain agnostic about this issue and show results with an without a mass cut. Figure 7 shows distributions of several template-inspired observables obtained from both the PYTHIA and SHERPA data. Since our focus is on the difference in the shapes of various observables, all of the kinematic distributions are shown after cuts slightly different from the ones of eq. (3.7), with $2t_b$ denoting that both two-body template momenta are *b*-tagged. Ref. [23] showed that at very high p_T (say above the TeV scale), $\bar{\theta}$ displays a sizable rejection power. Our result shows that at lower p_T , in the region where small Rapproximation does not hold, the background discriminating power of $\bar{\theta}$ is highly diminished. In addition to Ov_3 , tPf and especially $\Delta R_{bb}/\Delta R_t$ appear to be promising variables. We discuss tPf in more detail in appendix A.

3.1.1 Background rejection power at $\sqrt{s} = 8$ TeV

We proceed to discuss the rejection power of the method for jets with $p_T > 300$ GeV at $\sqrt{s} = 8$ TeV. We define the signal efficiency ϵ_s to be the ratio of the signal cross section after various kinematic cuts are applied, to the cross section after Basic Cuts, i.e.

$$\epsilon_s \equiv \frac{\sigma_{Wh}^{cuts}}{\sigma_{Wh}^{BC}} \tag{3.3}$$

where *BC* denotes Basic Cuts of eq. 3.2 and *cuts* are any kinematic cuts in addition to the Basic Cuts (see eq. (3.7)). Each cross section in eq. (3.3) assumes *b*-tagging efficiency of 75%, *c* mistag rate of 18% and light jet mistag rate of 1%.

Similarly, we define the *total* background fake rate to be

$$\epsilon_f \equiv \frac{\sigma_{Wb\bar{b}}^{cuts} + \sigma_{t\bar{t}}^{cuts}}{\sigma_{Wb\bar{b}}^{BC} + \sigma_{t\bar{t}}^{BC}},\tag{3.4}$$

and the fake rate for individual background channels as

$$\epsilon_{Wb\bar{b}} \equiv \frac{\sigma_{Wb\bar{b}}^{cuts}}{\sigma_{Wb\bar{b}}^{BC}}, \quad \epsilon_{t\bar{t}} \equiv \frac{\sigma_{t\bar{t}}^{cuts}}{\sigma_{t\bar{t}}^{BC}}.$$
(3.5)

Finally, we define the background rejection power RP as

$$RP \equiv \frac{\epsilon_s}{\epsilon_f}.$$
(3.6)

For the purpose of illustration, we consider several combinations of cuts on both template and jet observables in addition to Basic Cuts of eq. (3.2), while we leave Ov_3^{\min} a free parameter. We label the cuts as following:

Cuts 1 :

$$Ov_2 > 0.9.$$

 Cuts 2 :
 $Ov_2 > 0.9, 2t^b.$

 Cuts 3 :
 $Ov_2 > 0.9, 2t^b, \Delta R_{bb}/\Delta R_t < 1.0.$

 Cuts 4 :
 $Ov_2 > 0.9, 2t^b, \text{tPf} < 0.3.$

 Cuts 5 :
 $Ov_2 > 0.9, 2t^b, \Delta R_{bb}/\Delta R_t < 1.0.$

 110 GeV < $m < 130$ GeV,
 (3.7)

where $2t^b$ denotes that both two-particle peak template momenta are *b*-tagged.



Figure 7. Distributions of various substructure and template-based observables obtained from Pythia (left) and Sherpa (right). Basic Cuts are applied to all panels in addition to the cuts specified in the label. The symbol $2t_b$ is used to denote that both two-body template momenta are b-tagged.



Figure 8. Background rejection power of the template analysis. The figure shows the overall efficiency and fake rate with fixed cuts of eq. (3.7). A cut on Ov_3^{\min} runs along the curves. All efficiencies are relative to Basic Cuts of eq. (3.2).

Figure 8 summarizes the results. The left panel shows rejection power obtained from our analysis, with and without a mass window cut. The signal efficiency and fake rates are measured relative to the cross sections with Basic Cuts from table 1. The curves also include *b*-tagging efficiencies we discussed the previous section. Each curve represents fake rate as a function of signal efficiency with a set of fixed cuts on template observables, while a cut on Ov_3 in the range of (0, 1) runs along the curve. Our results show that template observables can significantly improve the background rejection power relative to Basic Cuts of eq. (3.2). Figure 9 illustrates the rejection power over individual background channels. Template Overlap method alone performs significantly better in rejecting $Wb\bar{b}$ events for most signal efficiencies, as shows in the left panel of figure 9. This is reasonable since $t\bar{t}$ events typically consist of two *b*-tagged jets and an additional fragment of a hadronically decaying *W* boson. Such a configuration is more likely to be tagged with a higher Ov_3 score than the typical two body substructure of a light QCD jet. However, cuts on additional kinematic observables such as the Template Stretch or tPf, as well heavy flavor tagging requirements, can result in $t\bar{t}$ events being rejected at a rate higher than $Wb\bar{b}$.

Table 2 shows an example of benchmark signal efficiency points. At 14% efficiency, a rejection factor of ≈ 4 is achievable with template based observables only. Additional mass window boosts the rejection power by $\approx 25\%$, leading to a $S/B \approx 0.3$, with roughly 20% efficiency.

Template Overlap Method can achieve enough rejection power at 8 TeV to overcome the backgrounds at the severe cost of signal efficiency. Yet, the ~ 20 fb⁻¹ of integrated luminosity collected during the LHC 8 TeV run is insufficient to yield a statistically significant Higgs signal in the boosted regime, primarily due to the tiny boosted signal cross section. For instance, take our results from table 2 with Cuts 5. The signal efficiency of 23% and 20 fb⁻¹ of data would yield 1 Higgs event with $S/\sqrt{B} \approx 0.45$. A 3σ signal would



Figure 9. Background rejection power of the template analysis for individual background channels. The left panel shows the overall efficiency and fake rate for $t\bar{t}$ and $Wb\bar{b}$ separately with Cuts 1 of eq. (3.7). A cut on Ov_3^{\min} runs along the curves. The right panel shows the signal efficiency and fake rate for Cuts 3 of eq. (3.7). All efficiencies are relative to Basic Cuts of eq. (3.2).

	Cut Set	Ov_3^{\min}	$\epsilon_s~(\%$)	σ_s (fb)	$\epsilon_{Wb\bar{b}}$	$\sigma_{Wb\bar{b}}$ (fb)	$\epsilon_{t\bar{t}}$	$\sigma_{t\bar{t}}$ (fb)	overall rejection power	S/B	$S/\sqrt{B} \left(20 \mathrm{fb}^{-1}\right)$
	Cuts 3	0.3	39.0	0.08	20.0	0.50	16.0	0.3	2.1	0.10	0.39
DVTHIA	Cuts 3	0.8	27.0	0.05	10.0	0.25	10.0	0.2	2.7	0.12	0.36
1 1 1 11174	Cuts 4	0.3	20.0	0.04	9.0	0.22	4.0	0.08	3.0	0.13	0.32
	Cuts 4	0.8	14.0	0.03	5.0	0.12	2.0	0.04	4.0	0.17	0.31
	Cuts 5	0.8	23.0	0.05	6.0	0.15	3.0	0.06	5.0	0.22	0.45
	Cut Set	Ov_3^{\min}	$\epsilon_s~(\%$)	σ_s (fb)	$\epsilon_{W b \bar b}$	$\sigma_{Wb\bar{b}}$ (fb)	$\epsilon_{t\bar{t}}$	$\sigma_{t\bar{t}}$ (fb)	overall rejection power	S/B	$S/\sqrt{B} \left(20 \mathrm{fb}^{-1}\right)$
	Cut Set Cuts 3	$\begin{array}{c} Ov_3^{\min} \\ 0.3 \end{array}$	$\frac{\epsilon_s~(\%~)}{39.0}$	$\frac{\sigma_s \text{ (fb)}}{0.08}$	$\epsilon_{Wb\bar{b}}$ 17.0	$\frac{\sigma_{Wb\bar{b}} \text{ (fb)}}{0.43}$	$\epsilon_{t\bar{t}}$	$\frac{\sigma_{t\bar{t}}~({\rm fb})}{-}$	overall rejection power 2.3	S/B 0.18	$\frac{S/\sqrt{B}(20{\rm fb}^{-1})}{0.54}$
SHEDDA	Cut Set Cuts 3 Cuts 3	$\begin{array}{c} Ov_3^{\min} \\ 0.3 \\ 0.8 \end{array}$	$\epsilon_s (\%)$ 39.0 26.0	$\sigma_s (fb)$ 0.08 0.05	$\begin{array}{c} \epsilon_{Wb\bar{b}} \\ 17.0 \\ 8.0 \end{array}$	$\sigma_{Wb\bar{b}}$ (fb) 0.43 0.20	$\epsilon_{t\bar{t}}$	$\sigma_{t\bar{t}} \text{ (fb)}$	overall rejection power 2.3 3.1	S/B 0.18 0.26	$ \frac{S/\sqrt{B} (20 \text{fb}^{-1})}{0.54} \\ 0.52 $
SHERPA	Cut Set Cuts 3 Cuts 3 Cuts 4	$\begin{array}{c} Ov_3^{\min} \\ 0.3 \\ 0.8 \\ 0.3 \end{array}$	$\epsilon_s (\%)$ 39.0 26.0 23.0	σ_s (fb) 0.08 0.05 0.05	$\epsilon_{Wb\bar{b}}$ 17.0 8.0 11.0	$ \begin{array}{c} \sigma_{Wb\bar{b}} \; ({\rm fb}) \\ 0.43 \\ 0.20 \\ 0.28 \end{array} $	ε _{tī} - -	$\sigma_{t\bar{t}} \text{ (fb)}$ $-$ $-$ $-$	overall rejection power 2.3 3.1 2.1	S/B 0.18 0.26 0.17	
SHERPA	Cut Set Cuts 3 Cuts 3 Cuts 4 Cuts 4	$\begin{array}{c} Ov_{3}^{\min} \\ 0.3 \\ 0.8 \\ 0.3 \\ 0.8 \end{array}$	$\epsilon_s (\%)$ 39.0 26.0 23.0 15.0	σ_s (fb) 0.08 0.05 0.05 0.03	$\epsilon_{Wb\bar{b}}$ 17.0 8.0 11.0 5.0	$ \begin{array}{c} \sigma_{Wb\bar{b}} \ ({\rm fb}) \\ 0.43 \\ 0.20 \\ 0.28 \\ 0.12 \end{array} $	€ _{tī} 	$\sigma_{t\bar{t}} \text{ (fb)}$ $-$ $-$ $-$ $-$	overall rejection power 2.3 3.1 2.1 2.9	S/B 0.18 0.26 0.17 0.24	$\frac{S/\sqrt{B} (20 \text{ fb}^{-1})}{0.54}$ 0.52 0.39 0.38
SHERPA	Cut Set Cuts 3 Cuts 3 Cuts 4 Cuts 4 Cuts 5	$\begin{array}{c} Ov_{3}^{\min} \\ 0.3 \\ 0.8 \\ 0.3 \\ 0.8 \\ 0.8 \\ 0.8 \end{array}$	$\begin{array}{c} \epsilon_s \ (\% \) \\ 39.0 \\ 26.0 \\ 23.0 \\ 15.0 \\ 17.0 \end{array}$		$\epsilon_{Wb\bar{b}}$ 17.0 8.0 11.0 5.0 4.0	$ \begin{array}{c} \sigma_{Wb\bar{b}} \ ({\rm fb}) \\ 0.43 \\ 0.20 \\ 0.28 \\ 0.12 \\ 0.10 \end{array} $	€tt̄ - - - -	σ _{tī} (fb)	overall rejection power 2.3 3.1 2.1 2.9 4.3	S/B 0.18 0.26 0.17 0.24 0.34	$\frac{S/\sqrt{B} (20 \text{fb}^{-1})}{0.54}$ 0.52 0.39 0.38 0.48

Table 2. Background Rejection Rates at $\sqrt{s} = 8$ TeV. The values in the table show the signal efficiencies (ϵ_s) and fake rates ($\epsilon_{Wb\bar{b}}, \epsilon_{t\bar{t}}$) relative to the cross sections with Basic Cuts of eq. (3.2) as well as the signal and background cross sections after the cuts specified in the first and second column. The overlap rejection power includes both the $Wb\bar{b}$ and $t\bar{t}$.

start appearing only with the un-realistic $\approx 1000 \, {\rm fb}^{-1}$ of data. We thus turn to projections for the future 13 TeV run.

3.1.2 Background rejection power at $\sqrt{s} = 13$ TeV

The composition of background channels at $\sqrt{s} = 13$ TeV changes relative to 8 TeV, with $t\bar{t}$ amounting to 60% of the total. Higher center of mass energy also allows us to push the minimum p_T to higher values; we opt for $p_T \geq 350$ GeV. Figure 10 shows the result. The jet cone scaling rule of eq. 3.1 with $p_T \geq 350$ GeV will always yield R = 0.8. We will thus use a fixed value of R = 0.8 in this section and neglect the scaling rule.

We again find that template overlap can significantly improve the rejection rate over traditional jet observables. The overall performance of templates improves relative to $\sqrt{s} = 8$ TeV and $p_T \ge 300$ GeV, as expected.



Figure 10. Background rejection power of the Template Overlap Method at $\sqrt{s} = 13$ TeV. The figure shows the overall efficiency and fake rate with fixed cuts of eq. (3.7). A cut on Ov_3^{\min} runs along the curves. All efficiencies are relative to Basic Cuts of eq. (3.2).

	Cut Set	Ov_3^{\min}	$\epsilon_s~(\%$)	σ_s (fb)	$\epsilon_{Wb\bar{b}}$	$\sigma_{W b \bar{b}} \ ({\rm fb})$	$\epsilon_{t\bar{t}}$	$\sigma_{t\bar{t}}~({\rm fb})$	overall rejection power	S/B	$S/\sqrt{B}(250{\rm fb}^{-1})$
	Cuts 3	0.3	35.2	0.1	19.1	0.3	11.8	0.4	2.4	0.16	2.0
PVTHIA	Cuts 3	0.8	27.0	0.08	10.5	0.2	10.0	0.3	3.2	0.17	1.9
I I I IIIA	Cuts 4	0.3	20.4	0.06	8.5	0.1	7.6	0.2	2.6	0.16	1.6
	Cuts 4	0.8	15.3	0.05	4.5	0.08	4.2	0.1	3.6	0.23	1.6
	Cuts 5	0.8	22.1	0.07	5.1	0.09	1.3	0.04	8.2	0.53	3.0

Table 3. Background Rejection Rates at $\sqrt{s} = 13$ TeV. The values in the table show the signal efficiencies (ϵ_s) and fake rates ($\epsilon_{Wb\bar{b}}, \epsilon_{t\bar{t}}$) relative to the cross sections with Basic Cuts of eq. (3.2) as well as the signal and background cross sections after the cuts specified in the first and second column. The Overall Rejection Power includes both the $Wb\bar{b}$ and $t\bar{t}$.

Figure 10 shows that an overall rejection power of ≈ 10 (Eff: 20%) is achievable at $\sqrt{s} = 13$ TeV, leading to an overall $S/B \approx 0.5$. This constitutes an improvement over the 8 TeV result where the maximum rejection power was ≈ 5 . Our results show that Template Overlap is capable of delivering a statistically significant boosted Higgs signal at $\sqrt{s} = 13$ TeV. At 22% signal efficiency (Cuts 5), we find that $S/B \approx 0.5$ is possible. The 3σ signal significance can be achieved with 250 fb⁻¹, the amount of data possible with the future LHC run.

3.2 Higgs tagging with Template Overlap - effects of pileup

A foe to most jet substructure observables, pileup has become an LHC fact of life. The strive for high luminosity resulted in pileup levels of whopping 20 average interactions per bunch crossing during the current 8 TeV run. Pileup events contribute both to the fat jet constituent multiplicity and the energy distribution within a jet, resulting in possibly dramatic effects on *any* jet substructure observable constructed out of jet constituents.



Figure 11. Jet mass distributions for signal and dominant backgrounds with pileup, $N_{\text{vtx}} = 20$. The left panels show distributions after the Basic Cuts of eq. (3.2), while the right panel shows the same distributions with $Ov_2 > 0.9 Ov_3 > 0.8$.

Jet mass is perhaps the best illustration of this point. Figure 11 shows an example. The left panel shows mass distributions in the presence of average 20 interactions per bunch crossing. In addition to shifting the mass peaks by as much as 30 GeV (relative to the plots in figure 6) pileup reduces the mass resolution. In a recent study of ref. [21], the ATLAS collaboration tested this feature with the LHC 7 TeV data. The mass resolution does not improve significantly even after cuts on the overlap are applied, again due to a low energy scale resolution. Note that reducing the value of σ_a and thus increasing the mass resolution of the Template Overlap Method could be used to study effects of pileup on jet observables.

Many "post processing" pileup subtraction techniques exist in the literature, such as trimming [18], pruning [19] and jet area techniques [29, 46]. Alternatively, ref. [47] presents a data driven method of correcting for pileup effects for jet shape variable of massive narrow jets. Finally, particle tracking information can be used to subtract pileup events, a method already used by the CMS collaboration [48]. In this section we do not consider any pileup subtraction. Instead, we show that the Template Overlap Method is largely unaffected by pileup.

Robustness of the Template Overlap Method against pileup comes from the definition of template overlap. Consider for instance a single template momentum p^t . The core of the overlap measure is the difference

$$\delta p_T = p_T^t - \sum_j p_T^j \times \theta(r_3 - \Delta R_{tj}), \qquad (3.8)$$

where p^{j} are momenta of jet constituents and θ selects the ones which fall into a cone of radius r_{3} around p^{t} . The size of the template subcone r_{3} thus limits the effects of pileup.

Matching p_T of a template state to the jet p_T with pileup is problematic. Jet p_T is shifted to higher values by pileup and as such is inappropriate as a criterion for template selection. Furthermore, a lower cut on the jet p_T with pileup will include jets which would not pass the cut if pileup was not present. We instead use the p_T of the leptonically



Figure 12. Comparison of boosted Higgs template results without and with pileup (20 average interactions per bunch crossing). Left panel shows a Higgs jet analyzed with no pileup. The right panel is the same jet with pileup added. Grey squares represent the jet constituents, the p_T of which is proportional to the size of the square. The solid circles are positions of *b*-quarks in the hard process. Particles with $p_T < 1$ GeV are not shown on the plot, but are included in the analysis.

decaying W boson, as an infra-red safe, pileup independent observable (recall that since the Higgs recoils agains a W boson, $p_W \approx p_H$, as we showed in section 3).

We simulate the effects of pileup by adding a random number of minimum bias events (MBE) to every event we analyze. The number of added MBEs is distributed according to a Poisson distribution with the mean $N_{\rm vtx} = 20$, consistent with the LHC conditions at $\sqrt{s} = 8$ TeV. Figure 12 shows an example of the pileup effects on the overlap analysis of a single event. The analysis of a jet with no pileup (left panel) yields nearly identical peak template state as the jet with pileup (right panel). Notice that the overlap scores remained within ~ 10% of each other. On distribution level, the situation is similar. Figure 13 shows examples of several template based observables with and without pileup. Even at 20 interactions per bunch crossing, we find no significant effects of pileup on the distribution shapes of Ov_3 , tPf and template stretch. In fact, the difference between susceptibility of tPf and Pf to a pileup environment is striking. While jet Planar Flow is significantly shifted to higher values by pileup, template Planar Flow remains mostly unaffected.

3.2.1 Rejection power

The overlap method used in this study effectively reduces the area of fat jets and is therefore less sensitive to pileup. As a simple application of this idea, in figure 14 we show fake rate vs. efficiency with the cuts of eq. (3.7), obtained from PYTHIA data at $\sqrt{s} = 8$ TeV and $N_{\rm vtx} = 20$. For a given cut, the efficiency is controlled by the lower cut on Ov_3 . The results depend on the choice of cuts, but it is clear that the overall performance of overlap approach remains largely unchanged when compared to the case of events without pileup. Our final results can be summarized in table 4 for a few benchmark efficiency points. We choose to omit the result of adding a mass cut since it requires an additional mechanism for pile up subtraction, and is thus beyond the scope of this paper. In each case, we find background rejections comparable to our results for events wihout pileup.



Figure 13. Effects of pileup on various jet-substructure distributions. Solid distributions were obtained with no pileup, while dashed distributions contain 20 average pileup events. All plots except the ones in the first row have a cut of $Ov_2 > 0.9$ for $r_2 = 0.3$.

	Cut Set	Ov_3^{\min}	Wh efficiency (%)	$W b \bar{b}$ fake rate (%)	$t\bar{t}$ fake rate (%)	overall rejection power
	Cuts 3	0.3	37.9	18.6	15.8	2.1
Pythia	Cuts 3	0.8	25.1	8.7	10.1	2.9
	Cuts 4	0.3	24.5	9.7	9.9	2.5
	Cuts 4	0.8	16.1	4.6	5.1	3.4

Table 4. Background Rejection Rates at $\sqrt{s} = 8$ TeV with $N_{xvt} = 20$. The values in the table show the signal efficiencies and fake rates relative to the cross sections with Basic Cuts. The overall rejection power includes both the $Wb\bar{b}$ and $t\bar{t}$.



Figure 14. Background rejection power of the Template Overlap Method at $\sqrt{s} = 8$ TeV with $N_{\text{vtx}} = 20$. The figure shows the overall efficiency and fake rate with fixed cuts of eq. (3.7). A cut on Ov_3^{\min} runs along the curves. All efficiencies are relative to Basic Cuts of eq. (3.2).

4 Conclusions

Hadronic decay channel of the Standard Model Higgs boson is one of the most challenging measurements in Higgs physics at the LHC. Traditional jet observables such as jet mass and p_T are inadequate to combat the large QCD background as well as high luminosity environments characteristic of the LHC. Jet substructure techniques can be used to overcome the large backgrounds in the boosted Higgs regime. Our results show that the template overlap method is not able to deliver sufficient background rejection power with a statistically significant signal in the realistic LHC conditions at $\sqrt{s} = 8$ TeV with 20 fb^{-1} . We also showed that one can achieve a good signal/background discrimination for a viable boosted Higgs search at $\sqrt{s} = 13$ TeV, starting with 250 fb^{-1} of data. We note that the above statement is based upon the assumption that the detector response would not result in a reduction of the efficiencies or in an increase of the fake rates estimated above. Detector effects (like calorimeter resolution or the smearing of the energy) were determined not to change qualitatively the phenomenological estimations in ref. [21], and we have neglected them in our analysis. A more complete study which makes use of the full detector simulation and reconstruction chain, including more realistic *b*-tagging, is beyond of the scope of this work, and we leave it to the experimental collaborations to study precisely how template-based variables are affected by detector response and systematic uncertainties.

In general the template-based variables presented here are robust against contamination from pileup at an average of 20 interactions per bunch crossing. We introduced several improvements into the template overlap framework. Varying the three-body subcones according to the transverse momentum of the template allows for more efficient signal tagging, while improving the background rejection power. Sequential template generation over longitudinally boost invariant variables results in a better phase space coverage at lower p_T (~ 200, 300 GeV), while keeping the number of templates at a reasonably low number of $O(10^6)$. For the first time, we integrated *b*-tagging identification into the peak templates. We further combined the *b*-tagging information with boosted event kinematics into template stretch, a new, useful observable. Template stretch is highly correlated with the jet mass, but with an important distinction that it is much more weakly affected by pileup. Finally, we improved on the existing definitions of Template Planar Flow, by including the simulated effects of parton showering via finite template subcones.

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A More on template Planar Flow

To illustrate how energy smearing affects the *parton* level tPf consider a peak three-body template state with an inertial tensor I_t . Energy smearing affects only the diagonal elements of I_t while off-diagonals remain unchanged due to symmetry, leading to

$$I_t \to I_s = I_t + \frac{P_T}{M} \, \alpha \, \mathbb{I},\tag{A.1}$$

where $P_T = \sum_{i=1}^{3} p_T^i$ and p_T^i is the transverse momentum of the *i*th template momentum. The effects of energy smearing are summarized in the symbol

$$\alpha = \int d\Omega \, g(\hat{p}_i),\tag{A.2}$$

where $g(\hat{p}_i)$ is the energy smearing distribution.

For simplicity, we consider a uniform distribution over a disc of radius r_3 , giving

$$\alpha_{cone} = \frac{r_3^2}{2}.\tag{A.3}$$

Alternatively, one could also consider a Gaussian distribution centered around the template momentum, with a width σ , resulting in

$$\alpha_{Gauss} = \sigma^2. \tag{A.4}$$

Continuing, the determinant of I_s becomes

$$det(I_s) = det\left(I_t + \frac{P_T}{M}\alpha\mathbb{I}\right)$$
$$= det\left(I_t\left(\mathbb{I} + I_t^{-1}\frac{P_T}{M}\alpha\mathbb{I}\right)\right)$$
$$= det(I_t)det\left(\mathbb{I} + I_t^{-1}\frac{P_T}{M}\alpha\mathbb{I}\right).$$
(A.5)

Similarly, the trace becomes

$$tr(I_s) = tr\left(I_t + \frac{P_T}{M}\alpha \mathbb{I}\right)$$
$$= Tr(I_t) + 2\frac{P_T}{M}\alpha$$
(A.6)

For a jet cone radius $R \sim 1$ it is reasonable to assume that $\alpha \ll 1$. Furthermore, P_T/M is typically of O(1), allowing us to expand $\det(I_s)$ in α . Keeping only the leading term in both the trace and the determinant we get

$$tPf_t \to tPf_s = tPf \times \frac{1 + tr(I_t^{-1})\frac{P_T\alpha}{M}}{1 + 4\frac{P_T\alpha}{Mtr(I_t)}}.$$
(A.7)

Since I_t is by definition a 2 × 2 matrix, we can write

$$\operatorname{tr}(I_t^{-1}) = \frac{\operatorname{tr}(I_t)}{\operatorname{det}(I_t)} = \frac{4}{\operatorname{tPf} \times \operatorname{tr}(I_t)}.$$
(A.8)

Combining eqs. (A.7) and (A.8) we finally obtain

$$tPf_s = tPf \times \frac{1 + \gamma/tPf}{1 + \gamma},$$
 (A.9)

where

$$\gamma \equiv \frac{4}{\operatorname{tr}(I_t)} \frac{P_T \alpha}{M}.$$
(A.10)

As $r_3 \to 0$, eq. (A.11) correctly reduces to the expression for tPf. In the limit of tPf $\to 1$, the showering effects become irrelevant as well, keeping eq. (A.7) bounded in accordance with the definition of tPf. Notice that eq. (A.11) is not valid in the tPf $\to 0$ limit, as we made an assumption that det $(I_t) \neq 0$ during its derivation.



Figure 15. Effects of showering on Planar Flow of the template momenta. The blue solid curve shows the jet Pf distribution. The red solid curve shows Template Planar Flow (tPf) calculated using only the three momenta from the peak template. The green, dashed curve shows Planar Flow distribution of peak templates with a sub cone of $r_3 = 0.12$ taken into account as in eq. (A.1). The black, solid line is the result of varying subcone radii according to the scaling rule in figure 1.

Continuing, in the narrow angle approximation $tr(I_t) = M/p_T^t$, where p_T^t is the total template transverse momentum. We have verified numerically that this result is satisfactory even at R=1.4 We can then rewrite eq. (A.7) as

$$tPf_s = tPf \times \frac{1 + \gamma'/tPf}{1 + \gamma'},$$
 (A.11)

where

$$\gamma' \equiv \left(\frac{4 P_T p_T^t}{M^2}\right) \alpha. \tag{A.12}$$

eq. (A.11) provides a useful qualitative insight into the effects of showering on Planar Flow. Template momentum configurations with low Planar Flow are more sensitive to showering effects. This is important when considering Planar Flow of Higgs jets which are expected to peak at low values.

Figure 15 shows the results of a data analysis with partonic tPf, and template Planar Flow calculated a in eq. (A.1), for both varying (tPf) and fixed r_3 . Addition of subcones to the Planar Flow calculation produces results which match the physical jet distribution much better than the parton level tPf(r = 0). Notice that the match between tPf and jet Planar Flow is excellent in the region of Pf > 0.4, where perturbative expansions are expected to hold.

B Efficient generation of template libraries

Template Overlap Method is a systematic framework aimed to identify kinematic characteristics of an boosted jet. A typical template configuration consists of a model template, f, calculated in perturbation theory, which describes a "prong-like" shape of the underlying hard subprocess of a jet. Template construction typically employs prior, theoretical knowledge of the signal kinematics and dynamics, as well as possible experimental input.

The simplest template configurations are the ones describing the kinematics of twobody processes such as the decay of SM Higgs or W/Z bosons into quark-antiquark pairs. These are easily dealt with by assuming the rest frame of the parent particle and producing two decay products with equal and opposite, isotropically-selected momenta and magnitude, subject to energy conservation. The problem of a N-body decay subtracts four constraints from the decay products' 3N degrees of freedom: three for overall conservation of momentum and one for energy.⁴ The final states can therefore be found on a (3N - 4)dimensional manifold in the multi-particle phase space. Note that the dimensionality of the template space increases rapidly with additional patrons. For instance, the two-body templates require only two degrees of freedom, while a corresponding four-body template space is already eight dimensional.

The question of which kinematic frame the templates should be generated in requires careful consideration. Authors of ref. [23] argued that a search for the global maximum of Ov_N could be too computationally intensive. To improve the computation time, the template states were generated in the Higgs rest frame using a Monte Carlo routine, and then boosted into the lab frame. While this method worked sufficiently well for tagging a highly-boosted object (i.e. a 1 TeV Higgs jet), it introduced residual algorithmic dependence and a certain sense of arbitrariness in the jet shape. At lower p_T the Monte Carlo approach samples mainly the templates within the soft-collinear region, leaving other regions of phase space unpopulated. An enormous number of templates is required to adequately cover the phase space at $p_T \sim O(100 \text{ GeV})$, thus fully diminishing the motivation for a Monte-Carlo approach. The simplest and most robust choice is then to generate templates directly in the lab frame and then rotate them into the frame of the jet axis. The result is a well covered template phase space in all relevant boosted frames. In addition, the lab frame templates result in a significant decrease in computation time as a much smaller number of templates are needed.

We proceed to show how to generate the phase space for 2- and 3-parton final states as well as how to generalize the results to arbitrary N.

B.1 The case of 2-body templates

First, we summarize our notation and conventions. The model template consists of a set of four vectors, p_1, \dots, p_N , on the hyperplane determined by the energy-momentum conservation,

$$\sum_{i} p_i = P, \ P^2 = M^2, \tag{B.1}$$

⁴For our purpose, a template is an object with no other properties other than its four-momenta.

where M, P are the mass and four momentum of a heavy boosted particle, i.e. the Higgs. For simplicity, we treat all template particles to be massless. We work in an (η, ϕ, p_T) space, where η is pseudorapidity, ϕ azimuthal angle and p_T transverse momentum. Without loss of generality, we can assume that the template points in the x direction $(\eta = \phi = 0)$. The templates are distributed according to

$$p_i = p_{T,i}(\cos\phi_i, \sin\phi_i, \sinh\eta_i, \cosh\eta_i), \ i = 1, 2, 3$$
(B.2)

subject to the constraint

$$\sum_{i=1}^{N} p_i = P = (p_T, 0, 0, E_J)$$
(B.3)

with $E_J = \sqrt{M^2 + p_T^2}$. We find it useful to define unit vectors by

$$\hat{p}_i = (\cos\phi_i, \sin\phi_i, \sinh\eta_i, \cosh\eta_i), \ i = 1, 2, \tag{B.4}$$

so that $p_i = p_{T,i} \hat{p}_i$.

Phase space for the 2-body decay processes is characterized by particularly simple kinematic parameters. To illustrate, first note that the 2-particle templates are uniquely determined by one single four momentum, p_1 subject to the condition

$$(P - p_1)^2 = 0. (B.5)$$

Writing $p_1 = p_{T,1}\hat{p}_1$, we can solve for $p_{T,1}$ in terms of the angles of the first parton

$$p_{T,1} = \frac{M^2}{2(P \cdot \hat{p}_1)}.$$
 (B.6)

We see that a 2-particle template is therefore completely determined in terms of the unit vector \hat{p}_1 as follows:

$$p_1 = \frac{M^2}{2(P \cdot \hat{p}_1)} \hat{p}_1 \tag{B.7}$$

$$p_2 = P - p_1.$$
 (B.8)

Note that we can represent such a template as a point $(\hat{\eta}, \hat{\phi})$ in $\eta - \phi$ plane. These are the two degrees of freedom, in accordance with the general result that the dimensionality of the N template space is 3N - 4.

B.2 The case of 3-body templates

A space of five degrees of freedom allows for 3-particle templates to differ from one another in more than one way. The 3-particle templates are determined by two four momenta, p_1 and p_2 , subject to the constraint,

$$(P - p_1 - p_2)^2 = 0. (B.9)$$

Using $p_1 = p_{T,1}\hat{p}_1$ and $p_2 = p_{T,2}\hat{p}_2$, we can solve for $p_{T,2}$ in terms of the angles of first two partons and $p_{T,1}$,

$$p_{T,2} = \frac{M^2 - 2P \cdot p_1}{2(P \cdot \hat{p}_2 - p_1 \cdot \hat{p}_2)}.$$
(B.10)

A general 3-particle template is then completely specified by $p_{T,1}$ and two unit vectors (or, equivalently, four angles) \hat{p}_1 and \hat{p}_2 .

B.3 Extension to arbitrary N

A generalization to an arbitrary number of particles is straight-forward. Proceeding as above, the N-particle templates are determined by p_1, \dots, p_{N-1} subject to the constraint,

$$\left(P - \sum_{i=1}^{N-1} p_i\right)^2 = 0.$$
(B.11)

Using $p_i = p_{T,i}\hat{p}_i$, we can now solve for $p_{T,N-1}$ in terms of the p_1, \dots, p_{N-2} and \hat{p}_{N-1} ,

$$p_{T,N-1} = \frac{M^2 + 2\sum_{i(B.12)$$

For the special cases of N = 2 and N = 3, this formula reduces to the above results .

B.4 Numerical simulations

We choose to cover the phase space uniformly in the (3N - 4) variables $\hat{p}_1, \dots, \hat{p}_{N-1}$ and $p_{T,1}, \dots, p_{T,N-2}$. For a finite number of points this method covers regions of phase space with large Planar Flow much more uniformly than a Monte Carlo based approach.

We survey the kinematically-allowed templates by fixing the total four momentum of each of the analyzed configurations and scanning over the possible values of $(p_{T,i})$ and the angles (\hat{p}_i) within the bounded interval. The number of variables depends on the number of degrees of freedom of the template states. The value of $P = (p_T, 0, 0, E_J)$ can be imposed by an additional equation:

$$p_N = P - \sum_{i=1}^{N-1} p_i.$$
(B.13)

We generate template libraries in the $\eta - \phi$ plane with a sequential scan in steps of $\Delta \eta = \Delta \phi = 0.05$. Three-particle templates we rehire and additional scan over the transverse momentum with which we perform in steps of $\Delta p_T = 5$ GeV. The resulting four momenta are a requirement that they "fit" into an anti-kT jet of fixed radius R. Templates with particles outside the jet cone are discarded. Our choice of step sizes leads to $O(10^4)$ 2-particle templates and $O(10^6)$ 3-particle templates.

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