Search for a Higgs boson decaying into a b-quark pair and produced in association with b quarks in proton–proton collisions at 7 TeV

CERN Collaboration

A search for a neutral Higgs boson decaying to a pair of b quarks, and produced in association with at least one additional b quark, is presented. Multijet final states with three jets identified as originating from b quarks, at least one of which may include a non-isolated muon, are studied. The data used in this analysis correspond to an integrated luminosity of 2.7–4.8 fb⁻¹, collected by the CMS experiment in proton–proton collisions at the LHC at a center-of-mass energy of 7 TeV. This search is particularly sensitive to Higgs bosons in scenarios of the Minimal Supersymmetric Model (MSSM) with large values of tan β. No excess over the predicted background from standard model processes is observed. Stringent upper limits on cross section times branching fraction are derived and interpreted as bounds in the MSSM parameter space. Observed 95% confidence level upper limits reach as low as tan β ≈ 18 for Mₐ ≈ 100 GeV.

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that include online algorithms for the identification of b jets to tackle the large multijet production rate at the LHC. The common analysis strategy is to search, in events identified as having at least three b jets, for a peak in the invariant mass distribution of the two leading b jets, i.e. those having the largest transverse momentum, over the large multijet background. A key point of both analyses is the estimation of the background using control data samples, which is addressed with different methods. The two analyses reach similar sensitivity to the MSSM Higgs scenarios described. The corresponding data sets are largely exclusive, and the small overlap is removed for the combined results.

2. The CMS experiment

The central feature of the CMS detector is a superconducting solenoid of 6 m internal diameter, providing a magnetic field of 3.8 T. Within the field volume, the inner tracker is formed by a silicon pixel and strip tracker. It measures charged particles within the pseudorapidity range \(|\eta| < 2.5\). The pseudorapidity is defined as \(\eta = -\ln(\tan(\theta/2))\) and \(\theta\) is the polar angle, while \(\phi\) is the azimuthal angle in radians. The tracker provides an impact parameter resolution of approximately 15 \(\mu\)m and a resolution on transverse momentum \((p_T)\) of about 1.5% for 100 GeV particles. Also inside the field volume are a crystal electromagnetic calorimeter and a brass/scintillator hadron calorimeter. Muons are measured in gas-ionization detectors embedded in the iron flux return yoke, in the pseudorapidity range \(|\eta| < 2.4\), with detector planes made using three technologies: drift tubes, cathode-strip chambers, and resistive-plate chambers. Matching muons to tracks measured in the silicon tracker results in a transverse momentum resolution between 1% and 5%, for \(p_T\) values up to 1 TeV. Extensive forward calorimetry complements the coverage provided by the barrel and endcap detectors. A more detailed description of the CMS detector can be found in Ref. [15].

3. Event reconstruction and simulation

The CMS particle-flow event reconstruction [16,17] is used for optimized reconstruction and identification of all particles in the event, i.e. electrons, muons, photons, charged hadrons, and neutral hadrons, with an extensive combination of all CMS detectors systems.

The reconstructed primary vertex with the largest \(p_T^2\)-sum of its associated tracks is selected and used as reference for the other physics objects.

Jets are reconstructed using the anti-\(k_t\) algorithm [18] from particle-flow objects with a radius parameter \(R = 0.5\) in the rapidity-azimuthal angle space. Each jet is required to have more than one track associated to it, and to have electromagnetic and hadronic energy fractions of at least 1% of the total jet energy. Additional proton–proton interactions within the same bunch crossing (pileup) affect the jet momentum reconstruction. To mitigate this effect, a track-based algorithm that removes all charged hadrons not originating from the primary interaction is used. In addition, a calorimeter-based algorithm evaluates the energy density in the calorimeter from interactions not related to the primary vertex, and subtracts it from the reconstructed jets in the event. Additional jet energy corrections [19] are applied.

Muons are reconstructed using both the inner silicon tracker and the outer muon system [20], and by performing a global track fit seeded by signals in the muon system.

The combined secondary vertex (CSV) algorithm [21] is used in the offline identification of b jets. The CSV algorithm uses information on track impact parameter and secondary vertices in a jet combined in a likelihood discriminant that provides a good separation between b jets and jets of other flavors. Secondary-vertex reconstruction is performed with an inclusive vertex search amongst the tracks associated to a jet [22].

Simulated samples of signal and background events were produced using various event generators and including pileup events. The CMS detector response is modeled with GEANT4 [23]. The MSSM Higgs signal samples, \(pp \rightarrow b\bar{b} \phi \rightarrow X, \phi \rightarrow b\bar{b}\), were produced with PYTHIA v6.424 [24], which yields the \(p_T\) and \(\eta\) distributions of the leading associated b jet in good agreement with the NLO calculations [25]. The Quantum Chromodynamics (QCD) multijet background events were produced with PYTHIA and ALPGEN [26], while for \(t\bar{t}\) + jets events the MADGRAPH [27] event generator was used. The next-to-leading order generators are interfaced with PYTHIA. For all generators, fragmentation, hadronization, and the underlying event are modeled using PYTHIA with tune Z2. The parton density functions (PDF) from CTEQ6L1 [28] used.

4. All-hadronic signature

We search for the Higgs boson in events where the three leading jets are all b-tagged. A signal would be identified as a peak in the invariant mass distribution of the two leading jets. Events in the data with only two b tags among the three leading jets are used to model the background, after proper reweighting, as described in Section 4.2.

4.1. Trigger and event selection

The large hadronic interaction rate at the LHC poses a major challenge for triggering. Events are accepted if either two or three jets are produced in the pseudorapidity range \(|\eta| < 2.6\) and have \(p_T\) above certain thresholds. Due to the increase in instantaneous luminosity as the run progressed the jet triggers had to be changed. Thus the data is divided into three categories. The first (second) category is characterized by dijet triggers in which the leading jet is required to have \(p_T > 46\) (60) GeV, and the next-to-leading jet \(p_T > 38\) (53) GeV. The third category is similar to the first but requires a third jet with \(p_T > 20\) GeV. The online identification of b jets is performed by an algorithm based on the impact parameter significance of the second most significant track associated to the jet as the b-tagging discriminant. Only events with at least two jets passing the online b-tagging requirement are accepted by the trigger.

The triggers with lower thresholds allow for a better exploration of the low-mass region, albeit with smaller integrated luminosity. The inclusion of the higher-threshold triggers allows higher integrated luminosity, but with the adjusted analysis requirements only the medium to high-mass region can be covered. For this reason two analysis scenarios are defined: in the low-mass scenario \((M_h < 180\) GeV), events accepted by the low jet \(p_T\) threshold triggers (first and third categories) are selected corresponding to an integrated luminosity of 2.7 fb\(^{-1}\). In the medium-mass scenario \((180 \leq M_h \leq 350\) GeV), a combination of dijet triggers with low and high jet \(p_T\) thresholds (first and second categories) forms an event sample with an integrated luminosity corresponding to 4.0 fb\(^{-1}\).

Events are required to have at least three reconstructed jets with \(|\eta| < 2.2\), where the b-tag efficiency and mistag probability are essentially constant. The three leading jets must also pass the \(p_T\) cuts of 46, 38 and 20 GeV (60, 53 and 20 GeV), respectively, in the low- (medium-) mass scenario. A minimal separation of \(\Delta R > 1\) between the two leading jets, where \(\Delta R = \sqrt{(\Delta \eta)^2 + (\Delta \phi)^2}\) and \(\Delta \eta\) and \(\Delta \phi\) are the pseudorapidity and azimuthal angle differences between the two jets, is required to suppress background from gluon splitting to a b-quark pair.
We define a “double-b-tag” sample to search for a signal by requiring all three leading jets to pass a tight CSV b-tagging selection requirement, consistent with the online b-tagging demand, at a working point characterized by a misidentification probability for light-flavor jets of about 0.1% at an average jet \( p_T \) of 80 GeV. The average b-tagging efficiency for true b jets is about 55% for jets with \( 80 < p_T < 120 \) GeV. The total numbers of events passing the trigger and offline selections are 106 626 and 89 637 for the low- and medium-mass scenarios, respectively. The efficiency of the trigger for signal events passing the offline selection is 47–67%, for a Higgs boson mass in the range of 90–350 GeV.

We define a “double-b-tag” sample, which is instrumental in estimating the shape of the background, where only two of the three leading jets have to pass the above-mentioned criteria, while the remaining untagged jet does not have to fulfill any b-tagging requirements. Since the double-b-tag sample is dominated by QCD events with two b jets, it represents a control region suitable to model the shape of the background contribution.

The secondary-vertex mass, namely the invariant mass calculated from all tracks forming the secondary vertex, provides an additional separation between b-, c- and light-flavor jets (attributed to u, d, s, or g partons) beyond the CSV b-tagging selection requirement. A compact b-tagging variable for the whole event is defined by assigning to each selected jet

\[ B_j = \min(B_j^{\text{CSV}}, B_j^{\text{CSV}}), \]

where \( B_j^{\text{CSV}} \) is set to zero. For intermediate values of the vertex mass between 1 and 2 GeV the index is set to 1, and for vertex mass larger than 2 GeV it is set to 2. The three indices \( B_1, B_2, \) and \( B_3 \) are combined in an event b-tagging variable \( X_{123} \), which is defined as follows:

\[ X_{123} = X_{12} + X_3, \]

where \( X_{12} = 0, 1 \) or 2 depending on whether \( B_1 + B_2 < 2, 2 \leq B_1 + B_2 < 3 \) or \( B_1 + B_2 \geq 3 \), respectively, and \( X_3 = 0 \) if \( B_3 < 2 \), and \( X_3 = 3 \) otherwise.

By construction, the event b-tag variable \( X_{123} \) can have six possible values ranging from 0 to 5. The intention of this mapping is to have each bin populated with sufficient statistics. For event types with a strong triple b-tag signature, the \( X_{123} \) distribution typically shows peaks at values of 2 and 5.

4.2. Background model and signal extraction

The dominant background comes from Quantum Chromodynamics (QCD) multijet production with two or more jets containing b hadrons, and can neither be fully reduced by kinematic selection, nor reliably predicted by Monte Carlo (MC) simulation. For this reason, a method based on control data samples, similar to the one used in Ref. [29], is applied. The background model is constructed from templates that are derived from the double-b-tag sample.

We divide the events in the double-b-tag sample into the following categories: bbX, bxb, and xbb, depending on the rank, sorted by \( p_T \), of the untagged jet, which is represented by the lower-case letter x. The ranking in descending \( p_T \) of the three jets is incorporated in the nomenclature adopted here, e.g. bbx means a sample of events where the two leading jets are b tagged and the third jet is the untagged jet. The true flavor of the untagged jet can be either light (u, d, s flavor quark, or g, denoted collectively by q), charm (c) or bottom (b).

From these three double-b-tag categories, nine background templates are constructed by weighting each untagged jet with the b-tagging probability assuming that its true flavor corresponds to either a light parton (u, d, s, or g, denoted by q), a charm (C) or a bottom (B) quark. The convention is that the capital letter indicates the assumed flavor of the untagged jet. The b-tagging probability for each flavor is determined as a function of jet \( p_T \) and \( \eta \) with simulated multijet events. Data/MC scale factors for the b-tagging efficiencies of b, c, and light-flavor jets are applied where appropriate [21].

Each background template is a distribution in the two-dimensional space spanned by \( M_{12} \), the dijet mass of the two leading jets, and the event b-tag variable \( X_{123} \).

The following nine background templates are thus created: Qbb; Cbb; Bbb; bQb; bCb; bbB; bbb; bbC; and bbB. In the bbB background events, two bb pairs are present. As pointed out in Ref. [29], the template bbB models mainly bb events in which the two leading b quarks originate from the same bb pair in the event, while bbB and bbB are important to cover cases where the two leading b quarks originate from different bb pairs.

The \( X_{123} \) dimension of the templates is modeled in a similar way. Each of the three possible values of the secondary-vertex mass index of the untagged jet is taken into account with a weight according to the probability that a jet will end up in a given bin of the secondary-vertex mass distribution. These probabilities, parametrized as a function of the jet \( p_T \) and \( \eta \), have been determined for each flavor using jets from simulated tt events.

Some of the nine templates are similar to each other in shape both for \( M_{12} \) and \( X_{123} \). In the cases where one of the two leading jets is untagged, e.g. Qbb, and bQb, the templates are combined, resulting in a merged template (Qb)b = Qbb + bQb. By analogy, also (Cb)b and (Bb)b are obtained. When the third-leading jet is the untagged one and the assumptions of its flavor are either Q or C, the bbQ and the bbC templates are combined to form the template bbX. The total number of templates to be fitted to the data is therefore reduced from nine to five, namely (Bb)b, (Cb)b, (Qb)b, bbB and bbX. The projections of the \( M_{12} \) and \( X_{123} \) variables are shown in Fig. 1 for the five background templates and for the low-mass scenario.

Templates whose dijet mass spectra resemble each other can be clearly distinguished with the introduction of the event b-tag variable \( X_{123} \). This is the case for example between (Bb)b and (Cb)b. In general, the event b-tag significantly improves the discrimination among all flavor components modeled.

The background templates whose projections are shown in Fig. 1 include two additional corrections. The basic assumption of the background model, that the double-b-tag sample (bb) consists entirely of events with at least two genuine b jets, is only approximately correct. Although the remaining contamination from non-bb events is indeed very small, the impact of the b-tagging selection could lead to distortions of the background model and a correction must be applied. This contamination is estimated directly from the data using a negative b-tagging discriminator [30] constructed with a track-counting algorithm based on the negative impact parameter of the tracks, ordered from the most negative impact parameter significance upward. The set of events in the double-b-tag sample in which at least one of the b-tagged jets passes a certain threshold of the negative b-tagging discriminator is used as a model for the contamination by non-bb events. The threshold is calibrated as a function of jet \( p_T \) with simulated multijet events, such that the negative tag rate equals the mistag rate. With this method, the non-bb contribution is found to be at the level of 3–4%. This correction results in only a marginal change in template shape. A second correction is necessary because the online b-tagging patterns differ in the double- and the triple-b-tag samples. The correction is determined from simulation, and is applied by appropriate weighting of the events in the double-b-tag sample.

A signal template is obtained for each considered value of the Higgs boson mass by performing the full selection on the events of the corresponding simulated signal sample. The mass resolution for combinations where both b jets stem from the Higgs decay
The expected background distribution and normalization is built using the same distribution for events with three jets of which only one is b-tagged, with a looser CSV b-tagging selection requirement, corresponding to a mistag probability of about 0.3%. The muon must be contained in one of the two leading jets. The final selection for the signal search adds the requirement that the third jet is also b-tagged, with a looser CSV b-tagging selection requirement. Two methods, both derived from data, have been developed to predict the expected background. The first is based on the computation of b-tagging probabilities of the third jet; and the second is based on a nearest-neighbor-in-parameter-space technique.

5.2. Background determination and signal extraction

As in the all-hadronic final state, the major backgrounds for the semileptonic final state are multijet events from hard-scattering processes. Other background processes, such as t + jets and Z → bb + jets, are predicted by the MC simulation to be less than 1% of the total background. Other possible backgrounds from events with multiple vector bosons (ZZ, ZW, WW) are negligible.

Two methods require a background-rich sample to serve as a control region. We construct a discriminating variable with a likelihood ratio, using various kinematic inputs: the \( p_T \) of the b jets; separation in \( \phi \) and \( \eta \) of the b jets; separation in \( \phi \) and \( \eta \) between the third jet and the combination of the two leading jets; and the b jet multiplicity. Two versions of this discriminating variable are used: one for the low-mass region \((M_b < 180 \text{ GeV})\) and another for the medium-mass region \((M_b > 180 \text{ GeV})\). For both mass ranges, the control region is defined as the sample of events having a low value for the discriminating variable, where the background is enriched and the signal depleted, and the signal region is defined as

5.1. Trigger and event selection

The data used in the semileptonic analysis were collected using different trigger selections, to cope with the increasing luminosity. All the triggers required a muon with a \( p_T > 12 \text{ GeV} \) threshold and the presence of one or two central jets \((|\eta| < 2.6)\) with transverse momentum above a given threshold \((20 \text{ or } 30 \text{ GeV}, \text{depending on the data-taking period})\). Furthermore, one or two b-tagged jets are required online. Initially the track with the second-most significant impact parameter was used. Later, when a second online b-tag was introduced, the selection was on the first track, in order to retain enough signal efficiency even with this tighter selection. An integrated luminosity of \(4.8 \text{ fb}^{-1}\) has been analyzed, and about \(1.67 \times 10^7\) events were collected.

The offline analysis requires a muon with \( p_T > 15 \text{ GeV} \), at least three jets with \(|\eta| < 2.6\), having transverse momentum \( p_T > 30 \text{ GeV} \) for the first two and \( p_T > 20 \text{ GeV} \) for the third one. The separation between any pair of jets has to be \(|\Delta R| > 1\). The two leading jets must be b-tagged using the CSV b-tagging algorithm with a working point giving mistag probability for light jets of about 0.3%. The muon must be contained in one of the two leading jets. The final selection for the signal search adds the requirement that the third jet is also b-tagged, with a looser CSV b-tagging selection requirement, corresponding to a mistag probability of about 1%. The total number of events which pass the selection is 60,195.

The relative efficiencies of the triggers with respect to the offline selection criteria were measured using lower-threshold single-muon triggers. These efficiencies are found to be about 45–60%, depending on the Higgs boson mass and the trigger.

5. Semileptonic signature

In the semileptonic signature, as for the all-hadronic one, a signal is searched for in events with three identified b jets, as a peak in the invariant mass distribution of the two leading jets. The expected background distribution and normalization is built using the same distribution for events with three jets of which only one or two are tagged as b jets, reweighting the events with a probability derived from a control region and computed with two different techniques. The muon requirement in the final state reduces the absolute signal efficiency, since it selects events where at least one of the b quarks decayed semileptonically in the muon channel, but it helps to reduce the event rate at the trigger level, allowing for a lower threshold for the jets.
the complementary sample. The number of events in the signal region is 33 366 and 16 866, for the low- and medium-mass regions, respectively.

The first method, henceforth called the matrix method, uses the probability that a jet is identified as a b jet to predict the background in the signal region. The predicted distribution of any observable $x$ in the three-b-jet sample can be calculated by rescaling, on an event-by-event basis, the same distribution for the $bbj$ sample by the probability $P_b$ of the third jet to be b-tagged. Taking into account the contribution of $b$, $c$, and light jets, the probability expands as: $P_b = \epsilon_b \cdot x_b + \epsilon_c \cdot x_c + \epsilon_q \cdot (1 - \epsilon_b - \epsilon_c)$, where $\epsilon_{b,c,q}$ are the probabilities (or b-tagging efficiencies) for a jet to be b-tagged, if it is originated from $b$, $c$, or light parton, respectively. The $\epsilon_{b,c}$ are the probabilities that the third jet originates from the corresponding quark, which depend on the kinematics of the event and of the third b jet.

The b-tagging efficiencies are taken from the MC simulation and checked with several methods derived from data [31]. Data/MC scale factors (close to unity within a few percent) are applied for the efficiencies and the corresponding uncertainties used as systematics uncertainties. The efficiencies are parametrized as functions of the third jet $p_T$, $\eta$ and charged-particle multiplicities. The quark flavor fractions are obtained directly from data by a simultaneous fit of two flavor-sensitive observables, using templates built from simulated events with $b$, $c$, and light quarks. The first variable used is a b-tag discriminator which uses the confidence level that the four tracks with the highest impact parameter in the jet are consistent with originating from the primary vertex. The second is the invariant mass associated with the secondary vertex, if it has been reconstructed. The parametrization for quark fractions also depends on the angular separation between the three b jets. Only events in the control region are used to obtain the quark fraction, which is then used to predict the background in both control and signal region.

The second method, called the nearest-neighbor method, exploits the fact that the probability for an event to appear signal-like depends on several event and jet variables. Events from the background enhanced control region are categorized according to several such variables, and are used to create a multi-parameter background prediction. The method uses the $bbj$ sample, and determines, for each event, the probability to pass the final selection. Starting from the $bbj$ sample, excluding the $bbj$ events, we can identify four disjoint subsets with which we work: (1) $bbj$ (including $bbj$) in the control region, (2) $bbj$ in the control region, (3) $bbj$ in the signal region, and (4) $bbj$ in the signal region. The sum of the above sets corresponds to the initial sample, where the leading jet is always b-tagged. We call collectively “training sample” the sum of subsets (1) and (2), and “testing sample” the sum of (3) and (4). The probability that an event in the testing sample passes the full selection is estimated by considering a larger sample of “similar” events in the training sample, and counting how many of these events pass the full selection.

For each event in the testing sample, referred to as test events, we select a sample of events with similar kinematics in the training sample. The probability for a test event to pass the final selection is calculated by selecting a sample of 100 training events inside a hyper-ellipsoid in the multi-dimensional space of event and jet observables, centered at the test event. The training events are chosen as those having the smallest multi-dimensional distance $D$, where $D^2 = \sum_{i=1}^{n_V} w_i^2(x_i^{\text{test}} - x_i^{\text{training}})^2$, $x_i$ is the variable defining the test and training event, and $n_V$ is the number of variables. The weights $w_i$ account for the different dispersions of the variables and their different sensitivity to the b-tagging probability.

A total of $n_V = 14$ different variables are used, including $p_T$, $\eta$ and the charged-particle multiplicity of the jets, the angular separation between the jets, and the invariant mass and transverse momentum of the combined jet–jet system. The weights $w_i$ are computed from the derivative of the probability for an event to pass the final selection as a function of the variable $x_i$. We then compute the numbers of $bbj$ and $bbj$ events inside this training sample. Finally, the probability for test events to have three b-tagged jets is computed as the ratio of $bbj$ to $bbj$ events in the training sample, using a weighted average, with $1/D^2$ as weight. The probability obtained this way is then applied, event-by-event, to the sample of
test events to predict the invariant mass distribution of the sample passing the final selection. This method gives a prediction of the background shape in the signal region independent from that obtained with the matrix method.

The background predictions for the invariant mass distribution of the two leading jets from the two methods described above are shown in Fig. 2. They are compared with the actual distribution in events with three b-tagged jets in the control region (low value of the discriminator), for low- and medium-mass regions. The predictions are normalized to the number of events seen; the absolute normalization of the prediction will be discussed in Section 6.

Because the matrix and nearest-neighbor methods use exclusive data samples, we can combine their results. This is done by performing a weighted average of their bin-by-bin predictions, using the statistical uncertainties $\sigma_i$ as weights ($w_i = 1/\sigma_i^2$). In case the $\chi^2$ of the average is greater than 1, $(\sqrt{\chi^2} - 1) \cdot \sigma_i$ is used, bin-by-bin, as an additional systematic uncertainty, following the Particle Data Group prescription [32].

### 6. Systematic uncertainties

Various systematic uncertainties on the expected signal and background estimates affect the cross section estimation and, consequently, its interpretation within the MSSM. In both analyses the main source of systematic uncertainty on the estimated signal yield comes from uncertainties related to jet reconstruction and b tagging. The second source is the turn-on behavior of the trigger efficiency, given the rather low thresholds used in the event selection. Other sources include uncertainties on the integrated luminosity and lepton identification. The theoretical cross sections used for the MSSM interpretation are subject to factorization and renormalization scale uncertainties, uncertainties due to the choice of parton distribution functions and $\alpha_s$, and uncertainties from the underlying event and parton shower modeling [33]. These uncertainties affect only the computation of the upper limits for the MSSM parameter $\tan \beta$ from the cross section results. The systematic effects directly affecting the signal efficiency, hence the cross section and MSSM interpretation, are summarized in Table 1.

There are systematic uncertainties that affect only the all-hadronic or semileptonic analyses. In the all-hadronic analysis, Table 1 includes systematic uncertainties related to the efficiency of the online b-tag selection relative to that applied offline, and to a slight dependence of the b-tagging efficiency on the jet topology. Various uncertainties also affect the shapes of the signal and background templates used in the fit. Shape-altering effects from uncertainties on the jet energy scale, jet energy resolution, b-tagging efficiency and mistag rates are accounted for in the fits with nuisance parameters. For the background templates, only the latter two are relevant. In the following we quantify background-related systematic uncertainties by their effect on the estimated signal fraction $f_{\text{sig}}$ (defined in Section 4). The uncertainty arising from the jet energy scale and the b-tagging efficiency on the template shape increases the $f_{\text{sig}}$ uncertainty by typically 0.1–0.4%; the corresponding effect from the jet energy resolution uncertainty is 0.1–0.3%. Additional shape-altering systematic uncertainties arise from the impurity of the double-b-tag sample and the online b-tagging correction to the background templates shape. The contribution of the former to the $f_{\text{sig}}$ systematic uncertainty ranges between 0.1% and 0.3% in the mass range 90–130 GeV, and is below 0.1% elsewhere. The effect of the latter correction ranges from 0.1% to 0.4% in the mass range 90–160 GeV, and is below 0.1% elsewhere.

In the semileptonic analysis there are uncertainties on both the background shape and normalization. The shape-related uncertainty is inferred by the comparison of the background predictions obtained with the two methods described in Section 5.2. The corresponding uncertainty scaling is included on a bin-by-bin basis in the binned maximum likelihood fit to the distribution of the final observable. The background normalization uncertainty has two components: the first is related to the level of agreement between the predicted $M_{12}$ distribution and the actual $b\bar{b}b$ one in the data control region and the second is related to the extrapolation of this prediction from the control region to the signal region. The ratio between the predicted $M_{12}$ distribution and the actual $b\bar{b}b$ one in the control region as seen in the data is used to normalize the prediction in the signal region, and its uncertainty is used as a systematic uncertainty. The scale factor is $0.877 \pm 0.007$ for the low-mass region and $0.885 \pm 0.006$ for the medium-mass region. For the extrapolation from the control region to the signal region, the MC simulation shows a constant ratio between the predicted $M_{12}$ distribution and the actual $b\bar{b}b$ one in the signal region. The additional correction is $1.01 \pm 0.04$ and $1.02 \pm 0.05$ for the low- and medium-mass regions, respectively. The uncertainties on these corrections are used as systematic uncertainties for the background normalization: 4.4% and 5.0% for the low- and medium-mass ranges, respectively.

### Table 1

<table>
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<th>Source</th>
<th>All-hadronic</th>
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<th>Type</th>
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<td>Underlying event and parton showering</td>
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A range indicates the variation across the probed Higgs boson mass values. The source with † also affects the background, while those with * only affect the model-dependent results in the space of the MSSM parameters $M_0$ and $\tan \beta$. The sources labeled with "rate" affect only the total signal yield, those with "shape" also the shape of the signal.
7. Results

In the all-hadronic analysis, we first test the background-only hypothesis by performing a \( \chi^2 \) fit without including a signal template, but only a linear combination of the background templates, as described in Section 4.2. The coefficients \( f_{i}^{bgd} \) are free parameters, but are constrained to be positive. Results are shown in Fig. 3 (a)–(c) for the low- and medium-mass scenarios. The background model fits the data well within the uncertainty propagated from the templates (hatched area). According to the fit, the templates associated with production of three b jets provide the dominant contribution to the background.

Subsequently, a signal template is included together with the background templates in the fit, with its fraction \( f_{\text{sig}} \) also allowed to vary freely. The fit is performed for Higgs boson masses from 90 to 350 GeV. The fit for a Higgs boson mass of 200 GeV in the medium-mass scenario is illustrated in Fig. 3(d).

The semileptonic analysis uses a binned likelihood fit to the invariant mass distribution of the two leading jets in the event to extract a possible MSSM Higgs contribution. Two different background predictions are considered, for the low- and medium-mass regions, which are fitted separately. In the fit the shape and normalization of the background component are constrained through nuisance parameters as explained in Section 6. The predicted back-
Fig. 4. Results from the semileptonic analysis. Data (red) and predicted background (blue) in the signal region, for (top) low-mass range (used for $M_{\phi} \leq 180$ GeV) and (bottom) medium-mass range (used for $M_{\phi} > 180$ GeV); the expected signal for different $M_A$ and for $\tan \beta = 30$ in the $m_{\text{max}} h$ scenario, as described in the text, is also plotted. The difference between data and predicted background is also shown: the blue area represent the systematic and statistical uncertainties on the background prediction. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this Letter.)

Fig. 5. Observed and expected upper limits for the cross section times branching fraction at 95% CL, with linear (top) and logarithmic (bottom) scales, including statistical and systematic uncertainties for the combined all-hadronic and semileptonic results. One- and two-standard deviation ranges for the expected upper limit are also shown.

No significant deviation from background is observed in either analysis, and the CLs [34–37] criterion is used to combine both results and determine the 95% confidence level (CL) limit on the signal contribution in the data, using the RooStats [38] package. To avoid correlations, in the all-hadronic analysis the events common to the semileptonic case are removed from the triple-b-tag samples. The fractions of events removed in the all-hadronic data samples are 2.3% and 2.7% for the low- and medium-mass scenarios, respectively. The requirement of a muon in the semileptonic analysis and the harder kinematic selections of the all-hadronic analysis are responsible for such small overlap. Overlapping events in the simulated signal samples are also removed, although they
are found to have negligible effect on the shape of the signal templates.

Results are shown graphically in Fig. 5 in terms of cross section times branching fraction, and reported in Table 2. There is generally good agreement between the observed and expected upper limits within statistical errors, and no indication of a signal is seen. The observed upper limits range from about 312 pb at $M_b = 90$ GeV to about 4 pb at $M_b = 350$ GeV. The all-hadronic signature has a generally larger signal efficiency, but requires higher thresholds for jet energies, while the presence of a muon in the semileptonic signature allows for lower thresholds at the cost of lower signal efficiency. As a result, both signatures are comparable in sensitivity.

Fig. 6 presents the results in the MSSM framework as a function of the MSSM parameters $M_A$ and $\tan \beta$, combining the individual results of the two analyses, including all the statistical and systematic uncertainties as well as correlations. We use the MSSM $m_{h_\text{max}}$ benchmark scenario [39,40], which is designed to maximize the theoretical upper bound on $M_h$ for a given $\tan \beta$ and fixed $M_{\text{SUSY}}$. Even though its parameters are under tension with the latest experimental results [41], it is currently still the most suitable benchmark scenario to compare the sensitivity of different analyses channels. The definition of theory parameters in the $m_{h_\text{max}}$ benchmark scenario is the following: $M_{\text{SUSY}} = 1$ TeV; $X_t = 2M_{\text{SUSY}}$; $\mu = 200$ GeV; $M_3 = 800$ GeV; $M_2 = 200$ GeV; and $A_0 = A_t = A_b = 800$ GeV. Here, $M_{\text{SUSY}}$ denotes the common soft-SUSY-breaking squark mass of the third generation; $X_t = A_t - \mu/\tan \beta^2$ is the stop mixing parameter; $A_t$ and $A_b$ are the stop and bottom trilinear couplings, respectively; $\mu$ is the Higgsino mass parameter; $M_3$ is the gluino mass; and $M_2$ is the SU(2)-gaugino mass parameter. The value of $M_1$ is fixed via the unification relation $M_1 = (5/3)M_2 \sin \theta_W/\cos \theta_W$. The expected cross section and branching fraction, in pb, and on $\tan \beta$ in the $m_{h_\text{max}}$ benchmark scenario for two values of the parameter $\mu = \pm 200$ GeV, are presented in Table 2.

8. Summary and conclusions

We searched for a Higgs boson decaying into a pair of $b$ quarks, produced in association with one or more additional $b$-quark jets. We used data samples corresponding to an integrated luminosity of 2.7–4.8 fb$^{-1}$ collected in 2011 in proton–proton collisions at a center-of-mass energy of 7 TeV at the LHC. The data were collected with dedicated multijet triggers including $b$-tag selection, utilizing both all-hadronic and semileptonic event signatures.

The search was performed on a triple-$b$-tag sample, using the invariant mass of two leading jets as a discriminating variable, with a prediction of the multijet background using control data samples. The all-hadronic analysis makes use of a second discriminating variable, $X_{123}$, that reflects the heavy flavor content of the event.

No signal is observed above the SM background expectations, and 95% confidence level upper limits on the $pp \rightarrow b\phi + X$, $\phi \rightarrow b\bar{b}$ cross section times branching fraction are derived in the 90–350 GeV mass range. These results are interpreted, in the MSSM model and the $m_{h_\text{max}}$ scenario, in terms of bounds in the space of the parameters $M_A$ and $\tan \beta$. The 95% confidence level bound on $\tan \beta$ varies from about 18 to 42 in this Higgs boson mass range, thus excluding a region of parameter space previously unexplored for this final state.

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Fig. 6. Observed upper limits at 95% CL on $\tan \beta$ as a function of $M_A$, including the statistical and systematic uncertainties, in the $m_{\max}^\text{max}$ benchmark scenario, both for $\mu = +200$ GeV (top) and $\mu = -200$ GeV (bottom), for the combined all-hadronic and semileptonic results. One- and two-standard deviation ranges for the expected upper limit are represented by the color bands. The expected upper limits for each of two signatures are also shown (dashed and dotted lines). (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this Letter.)

Fig. 7. Observed upper limits at 95% CL on $\tan \beta$ as a function of $M_A$, including the statistical and systematic uncertainties, in the $m_{\max}^\text{max}$ benchmark scenario with $\mu = -200$ GeV for the combined all-hadronic and semileptonic results. One- and two-standard deviation ranges for the expected upper limit are represented by the gray bands. Previous exclusion regions from LEP [8] and Tevatron in the multi-b jet channel [12] are overlaid.

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