

Identification and tagging of double b -hadron
jets with the ATLAS Detector

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Doctoral Thesis in Physics

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University of Buenos Aires

November 2012



UNIVERSIDAD DE BUENOS AIRES

Facultad de Ciencias Exactas y Naturales

Departamento de Física

Identificación y etiquetado de jets con dos hadrones b con el detector ATLAS.

Trabajo de Tesis para optar por el título de
Doctor de la Universidad de Buenos Aires en el área Ciencias Físicas

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Buenos Aires, Noviembre 2012

Agradecimientos

Quiero agradecer a mi director, Ricardo Piegaia, por su enseñanza constante, dedicación y amistad y a todos aquellos que trabajaron junto conmigo en el experimento ATLAS, Gastoncito Romeo (mi compañero de aventuras desde el comienzo), Gustavo Otero y Garzón, Hernán Reisin y Sabrina Sacerdotti. Un especial agradecimiento a Ariel Schwartzman y su equipo por las ideas y el soporte técnico constante. Quiero agradecer al Laboratorio CERN, al Experimento ATLAS, al programa HELEN y al programa e-Planet por permitirme ser parte de un proyecto increíble.

Quiero agradecer también a mis compañeros de grupo y oficina, Javier Tiffenberg, Yann Guardincerri, Pablo Pieroni y Orel Gueta por estar siempre dispuestos a darme una mano. Quiero agradecer el apoyo de mis compañeros de la carrera, especialmente a mis amigos Cecilia Bejarano y Tomas Teitelbaum. Quiero agradecer a los amigos que hice a lo largo de estos años en mis visitas al Laboratorio CERN, y a mis colegas y amigos de la Universidad de la Plata. Un especial agradecimiento a Fernando Monticelli.

Quiero agradecer a mis amigos de la vida por continuar a mi lado a pesar de las ausencias. Agradezco profundamente a toda mi familia por su apoyo y comprensión, especialmente a Cristina Silva, Lorena González y Juan Martín Alba.

Finalmente, quiero agradecer al CONICET y a la Fundación Exactas por hacer posible la realización de esta tesis.

Identificación y etiquetado de jets con dos hadrones b con el detector ATLAS

Resumen

Esta tesis describe un método que permite la identificación de jets que contienen dos hadrones b , que se originan en la división de un gluon en un par $b\bar{b}$. La técnica desarrollada explota las diferencias cinemáticas entre los llamados jets “merged” y los genuinos jets b , usando variables que describen la estructura interna y la forma de los jets, construídas a partir de las trazas asociadas a los mismos. Las variables con mayor poder discriminador son combinadas en un análisis de multivariable. Poder identificar y remover jets b que provienen de la división de un gluon es importante para la estimación y la reducción del fondo a señales de física dentro del Modelo Estándar y en nueva física. El algoritmo diseñado rechaza, en eventos simulados, el 95% (50%) de los jets “merged”, mientras que retiene el 50% (90%) de los jets b genuinos.

Palabras clave: Experimento ATLAS, Jets, Subestructura de Jets, QCD, Producción de jets b , Etiquetado de Jets b .

Identification and tagging of double b -hadron jets with the ATLAS Detector

Abstract

This thesis describes a method that allows the identification of double b -hadron jets originating from gluon-splitting. The technique exploits the kinematic differences between the so called “merged” jets and single b -hadron jets using track-based jet shape and jet substructure variables combined in a multivariate likelihood analysis. The ability to reject b -jets from gluon splitting is important to reduce and to improve the estimation of the b -tag background in Standard Model analyses and in new physics searches involving b -jets in the final state. In the simulation, the algorithm rejects 95% (50%) of merged jets while retaining 50% (90%) of the tagged b -jets, although the exact values depend on the jet p_T .

Keywords: ATLAS Experiment, Jets, Jet Substructure, b -jet Production, QCD, Gluon Splitting, b -tagging.

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Chapter 1

Introduction

The first years of proton-proton collisions at a centre of mass energy of 7 TeV delivered by the Large Hadron Collider and recorded by the ATLAS experiment have provided data to explore quantum chromodynamics (QCD) at scales never reached before. Precision measurements of strong interactions are interesting in their own right, but, in addition, QCD provides one of the main backgrounds to many New Physics measurements; furthermore, it is also through tests of QCD that New Physics may be discovered.

Due to QCD confinement the experimental signature of quarks and gluons are not the quarks and gluons themselves but a spray of “colorless” hadrons, that we call *jets*. Hadronic jets are a fundamental ingredient for precision tests of QCD: understanding and measuring their performance is crucial in the LHC environment. A wide range of physics signatures, within the Standard Model (SM) and Beyond the Standard Model (BSM) predictions, contain jets originating from bottom (b) quarks. The ability to identify jets containing b -hadrons, the product of the hadronization of b -quarks, is therefore important for the high- p_T physics program of the ATLAS experiment.

b -tagging algorithms rely on the relatively long decay length of b -hadrons

that gives rise to large impact parameter tracks and displaced decay secondary vertices; or on the presence of a soft lepton within the jet, the product of the semileptonic b -decay. These algorithms, however, do not provide information on the number of b -hadrons within the jet. In particular, they tag “merged” jets containing a $b\bar{b}$ pair, with no net heavy flavour, which do not correspond to the intuitive picture of a b -jet as a jet containing a single b -quark or antiquark.

The ability to single out merged b -jets has several applications. The measurement of the QCD bottom production if of great importance due to the correspondence between parton level production and the observed hadron level, and its potential to provide information on the b -quark parton distribution function. The theoretical calculation of the inclusive b -jet spectrum presents rather large uncertainties ($\sim 50\%$), considerably larger than those for the light jet inclusive spectrum ($\sim 10 - 20\%$) [1]. It is found that the largest uncertainties are associated to the production channel known as “gluon splitting” (GSP), where a gluon from the hard scatter decays into a close-by $b\bar{b}$, that a jet clustering algorithm often classifies within the same jet. This channel receives a strong enhancement from collinear logarithms spoiling the convergence of the perturbative series. An improvement in the accuracy of the theoretical predictions could be achieved by not including in the production cross-section the contribution from double b -hadron jets, which in QCD are produced $\sim 95\%$ of the time by the GSP channel.

Efficient tagging of merged b -jets can provide an important handle to understand, estimate and/or reject b -tagged backgrounds to SM analyses at the LHC that rely on the presence of single b -jets in the final state, such as top quark physics (either in the $t\bar{t}$ or the single top channels) or associated Higgs production ($WH \rightarrow \ell\nu b\bar{b}$ and $ZH \rightarrow \nu\nu b\bar{b}$). These processes suffer from

backgrounds that can be in part removed by a merged b -jet tagger. Jets containing a single b -quark or antiquark also enter in many BSM collider searches, notably because b -quarks are produced in the decays both of heavy SM particles (top quarks, the Z boson and the Higgs boson), and of particles appearing in proposed extensions of the SM. The ability to distinguish single b -jets from jets containing two b -hadrons is thus here of wide application to reduce SM backgrounds giving rise to close-by $b\bar{b}$ pairs.

At the Tevatron acelerator (Fermilab) 50% of the b -hadrons are due to the gluon splitting process; a larger fraction is expected to contribute at the LHC. There are two possible strategies to attempt to identify b -jets containing two b -hadrons in hadronic collisions. One of them relies on the direct reconstruction of the two b -decay secondary vertices [2]. This allows the measurement of the angular separation between the b -hadrons, but suffers from the low efficiency of a double b -tag requirement plus additional reconstruction inefficiencies at small angular separation between the two b -hadrons. In this thesis we develop for the first time in ATLAS an alternative method that does not rely on explicit vertex finding, but exploits the substructure differences between single and merged b -jets, combining them in a multivariate analysis.

The thesis is organized as follows: Chapter 2 describes the theoretical framework, with emphasis in the theory of the strong interactions and the aspects that are important for the understanding of the hadronic final state in hadronic collisions. The LHC and the ATLAS detector components are described in Chapter 3, together with a summary of the detector conditions during 2011 data taking. Chapter 4 details how jet reconstruction and calibration are performed at ATLAS and describes the procedure for the identification of b -quark jets. Chapter 5 presents the analysis of jet shape

and substructure variables for the discrimination between single and double b -hadron jets. The validation of the variables in 2011 data is also included. The construction of the multivariate discriminator and the discussion of the systematic uncertainties are presented in Chapter 6. Finally, chapter 7 summarizes the results.

Chapter 2

Theoretical framework

In this chapter a short overview of the theory of elementary particles and fundamental interactions is presented, with emphasis on the strong interactions and the description of the hadronic final state in hadron collisions.

2.1 The Standard Model

The Standard Model (SM) is a quantum field theory that describes the behavior of all experimentally-observed particles under the influence of the electromagnetic, weak and strong forces¹. In this model, all forces of nature are the result of particle exchange. The force mediators interact on the particles of matter, and, in some cases, due to the non-Abelian character of the theory², with each other.

The fundamental building blocks of matter predicted by the theory are

¹In principle gravitational forces should also be included in the list of fundamental interactions but their impact is fortunately negligible at the distance and energy scales usually considered in particle physics experiments.

²The transformations of the symmetry group do not commute in the case of the QCD and weak groups.

fermions with spin 1/2:

- six leptons (and their antiparticles), organized in three families,

$$\begin{pmatrix} \nu_e \\ e^- \end{pmatrix} \begin{pmatrix} \nu_\mu \\ \mu^- \end{pmatrix} \begin{pmatrix} \nu_\tau \\ \tau^- \end{pmatrix}$$

- and six quarks (and their antiparticles), organized in three families,

$$\begin{pmatrix} u \\ d \end{pmatrix} \begin{pmatrix} c \\ s \end{pmatrix} \begin{pmatrix} t \\ b \end{pmatrix}$$

The six types of quark are also known as the six quark flavors. Collectively, the u (up), d (down), and s (strange) quarks are frequently referred

to as the light quarks. The heaviest quark of the Standard Model, the quark t (top), was the last to be found [3, 4]. These particles are considered point-like, as there is no evidence of any internal structure of leptons or quarks to date.

In addition, the model contains the vector bosons which are the carriers of the fundamental forces:

- a gauge boson for the electromagnetic interactions, the photon γ ;
- three gauges bosons for the weak interactions, W^\pm and Z^0 ;
- eight gauge bosons for the strong interactions, called gluons.

The exact symmetry (see below) of the SM predicts massless particles. One possible mechanism for breaking this symmetry is the existence of a massive scalar Higgs field that has non-zero vacuum expectation value [5].

Very recently, a Higgs-like particle was discovered by ATLAS and CMS experiments at the LHC [6]. This scalar boson completes the table of Standard Model particles.

Gauge invariance, defined as the invariance of the theory under local transformations, is a fundamental property of the SM. In the theory, electromagnetism (Quantum electrodynamics), the weak interaction, and the strong nuclear force (Quantum Chromodynamics) are all derived from imposing Lorentz invariant symmetries onto the interacting fields.

The theory of Quantum electrodynamics (QED) describes the interaction of charged particles via the exchange of one (or more) photon. It is formulated by imposing a $U(1)$ or rotational symmetry onto the simplest field lagrangian that obeys the correct equation of motion. The full theory of QED was developed by Richard Feynman and others throughout the 1940s [7]. The structure of the SM is in a sense a generalisation of this theory, extending the gauge invariance of electrodynamics to a larger set of conserved currents and charges.

The symmetry associated to the weak interaction is the $SU(2)$ symmetry, which corresponds to rotations of 2-dimensional vectors. The latter combines with the $U(1)$ symmetry from QED to produce additional gauge fields. The gauge fields merge with the gauge field from QED to form W^+ , W^- and Z^0 bosons that are the carriers of the weak force. Unlike the photon, which is massless, the W^\pm and Z^0 bosons have masses close to 80 and 90 GeV, respectively [8, 9]. Due to these large masses, the weak force has a short range and is feeble at low energies. At masses higher than the Z mass, the electromagnetic and weak forces unify into a single force, known as the electroweak force [10, 11, 12].

The current theoretical theory of the strong interactions began with the

identification of the elementary fermions that make up the proton and other hadrons. In 1963, Gell-Mann and Zweig proposed the quark model [13, 14, 15], which asserts that these particles are in fact composites of smaller constituents. Mesons were expected to be quark-antiquark bound states and Baryons were interpreted as bound states of three quarks, all with fractional charges. The quark model was formalized into the theory of Quantum Chromodynamics (QCD) by Harold Fritzsch and Murray Gell-Mann [16] in 1973, who proposed that quarks carried an additional quantum number called color. Without color charge, it would seem that the quarks inside some hadrons exist in identical quantum states, in violation of the Pauli exclusion principle (this was indeed the problem of the quark model as proposed by Gell-Mann and Zweig). The color theory extends the electroweak Lagrangian to be symmetric under $SU(3)$ transformations, which introduces eight new physical gauge fields, the gluons. Due to Richard Feynman's parton model nomenclature [17], both quarks and gluons are commonly referred to as partons.

Another problem of the quark model was that free particles with fractional charges were never found. The answer to why we never see free quarks or gluons outside of a hadron, together with the tools for performing theoretical calculations in QCD are given the next sections.

2.2 Perturbative QCD

As described above, the fundamental actors of the theory of the strong interactions are quarks and gluons or, collectively, partons. Partons are confined in hadrons, but, act free at sufficiently small scales. This behaviour is called asymptotic freedom. The essence of asymptotic freedom is that the strong force couples particles together more strongly as the distance between

them increases. The experimental consequence of asymptotic freedom is that quarks and gluons require interactions with high energy probes to be ejected from nucleons, and they cannot be observed directly.

First indications of the presence of quarks resulted from the measurement of deep inelastic lepton-hadron scattering. The momentum transfer, Q^2 , between the probe particles (leptons) and the target hadron is analogous to the distance scale within the hadron being measured. The variation of the strength of the coupling with the energy is referred to as the “running” of coupling constant.

The low value of the strong coupling constant at high-energies permits the use of perturbative techniques to calculate physical processes. As one goes higher in the perturbative expansion, each term contains an additional factor of the coupling constant, α_s . Since α_s depends on the energy, it must be evaluated at some energy scale, close to the energy scale involved in the process. For instance, at an energy of 15 GeV, the strong coupling constant takes on an approximate value of 0.1, thus, from an expansion of an infinite number of terms, only a few need to be computed. The complexity of the process determines the precision of the calculation that can be performed. For example, predictions for the cross section for events with three partons in the final state are only available up to leading-order (LO). For inclusive parton production, calculations are typically performed at next-to-leading order (NLO). Feynman diagrams are used in the computation of the multiple terms in the expansion, they are graphical representations of each term.

With this formalism, the cross-section for the interaction of partons can be computed up to fixed-order in perturbation theory. In the case of hadron colliders, such as the LHC, the factorization theorem [18] allows the perturbative calculations for parton interactions to be extended to proton-proton

collisions.

In the simplest picture, a proton is a combination of three quarks: two up quarks and one down quark. The reality, however, is much more complex. In the proton there are gluons constantly being emitted and absorbed, causing quark/antiquark pairs of many flavors to be briefly produced and destroyed. The up and down quarks of the standard hadron model are called valence quarks, while the virtual quark/antiquark pairs are known as sea quarks. Both valence quarks and sea quarks, along with the gluons, share the total momentum of the hadron. The distribution of the momentum fraction, x , carried by each parton is expressed as a probability to find a particular parton with a given x . The latter is known as the Parton Distribution Function (PDF).

For cross-section calculations, the PDFs are evaluated at a factorization scale, μ_f , which can be thought of as the scale that separates short-distance, perturbative physics from long-distance, non perturbative physics. Any variation in the computed cross sections due to different choices of the energy scales can be interpreted as an uncertainty due to the unknown higher-order corrections in the cross-section calculation.

The evolution of the PDF on Q is given by the DGLAP equations, published separately in the 1970s by Yuri Dokshitzer, Vladimir Gribov and Lev Lipatov, and Guido Altarelli and Giorgio Parisi [19]. The DGLAP equations are derived by noting that the cross-section of a process should be independent of the factorization scale μ_f . This gives a precise mathematical form to the dependence. The dependence on x , however, must be obtained by fitting possible cross section predictions to data from hard scattering experiments.

2.3 Jet physics

Due to confinement the experimental signature of quarks and gluons are the final state “colorless” hadrons³. The packet of particles produced tends to travel collinearly with the direction of the initiator quark or gluon. The result is a collimated “spray” of hadrons (also photons and leptons) entering the detector in place of the original parton; these clusters of objects are what we define as jets. The first evidence for jet production was observed in e^+e^- collisions at the SPEAR storage ring at SLAC in 1975 [20].

The evolution from a single parton to an ensemble of hadrons occurs through the processes of parton showering and hadronization. Since the strong coupling constant grows with increasing distance between color charges, a strong color potential forms as the parton from the “hard” (high Q^2) scattering process separates from the original hadron. This large potential causes quark/antiquark pairs ($q\bar{q}$) to be created, each carrying some of the energy and momentum of the original partons. As these new partons move away from one another, yet more color potentials are formed, and the process repeats. Thus from one parton a shower of partons appears, traveling along the same direction as the original. This process continues until there is no longer enough energy to create additional $q\bar{q}$ pairs, and instead the remaining partons combine to form stable hadrons. Since this progression involves successively lower energies and lower momentum transfers, perturbative QCD cannot describe the full process. The full parton shower and hadronization process then cannot be calculated from first principles, but has to be modelled.

³We use “colorless” to mean a singlet representation of the color group.

2.3.1 Monte Carlo tools

Knowing QCD predictions is crucial in the design of methods to search for new physics, as well as for extracting meaning from data. Different techniques can be used to make QCD predictions at hadron colliders, and in particular at the LHC. The so called Matrix Element Monte Carlos use direct perturbative calculations of the cross-section matrix elements for each relevant partonic subprocesses. LO and NLO calculations are available for many processes. These “fixed-order predictions” include the first terms in the QCD perturbative expansion for a given cross-section; as more terms are involved in the expansion, an improvement in the accuracy of the prediction is expected. The complexity of the calculations increases significantly with the number of outgoing legs. Matrix element MC programs include ALPGEN [21], MADGRAPH [22] and others.

An alternative approach is applied by the so called Monte Carlo parton shower programs. These simulation programs use LO perturbative calculations of matrix elements for $2 \rightarrow 2$ processes, relying on the parton shower to produce the equivalent of multi-parton final state. PYTHIA [23] and HERWIG++ [24] are the most commonly used parton shower Monte Carlos.

The Monte Carlo generators must account for and correctly model the showering of partons. To approximate the energy-evolution of the shower, the DGLP equations that describe the evolution of the PDFs with changing energy scale can be used. The separation of radiation into initial- (before the hard scattering process takes place) and final-state showers is arbitrary, but sometimes convenient. In both initial- and final-state showers, the structure is given in terms of branchings $a \rightarrow bc$: $q \rightarrow qg$, $q \rightarrow q\gamma$, $g \rightarrow gg$ and $g \rightarrow q\bar{q}$. Parton b carries a fraction z of the energy of the mother energy and

parton c carries the remaining $1 - z$ (the term “partons” includes the radiated photons). In turn, daughters b and c may also branch, and so on. Each parton is characterized by some evolution scale, which gives an approximate sense of time ordering to the cascade. In the initial-state shower, the evolution scale values are gradually increasing as the hard scattering is approached, while these values decrease in the final-state showers. The evolution variable of the cascade in the case of PYTHIA, Q^2 , has traditionally been associated with the m^2 of the branching partons⁴. In the recent version of PYTHIA a p_\perp -ordered shower algorithm, with $Q^2 = p_\perp^2$ is available, and the shower evolution is cut off at some lower scale Q_0 typically around 1 GeV for QCD branchings. HERWIG++ provides a shower model which is angular-ordered.

There are two leading models for the description of the non-perturbative process of hadronization, after parton showering. PYTHIA uses the Lund string model of hadronization to form particles [25]. This model involves stretching a colour “string” across quarks and gluons and breaking it up into hadrons. HERWIG++ utilizes the cluster model of hadronization. In this model each gluon is split into a $q\bar{q}$ pair and then quarks and anti-quarks are grouped into colourless “clusters”, which then give the hadrons.

Hadronization models involve a number of non-perturbative parameters. The parton-shower itself involves the non-perturbative cut-off Q_0^2 . These different parameters are usually tuned to data from the LEP experiments.

In addition to the hard interaction that is generated by the Monte Carlo simulation, it is also necessary to account for the interactions between the incoming proton remnants. This is usually modelled through multiple extra

⁴The final-state partons have $m^2 > 0$. For initial-state showers the evolution variable is $Q^2 = -m^2$, which is required to be strictly increasing along the shower.

$2 \rightarrow 2$ scattering, occurring at a scale of a few GeV. This effect is known as multiple parton interactions (MPIs). In addition, these partons may radiate some of their energy, either before or after the hard interaction. All the additional parton interactions, which are not involved in the hard scattering process, are grouped together in the term underlying event. The modelling of the underlying event is crucial in order to give an accurate reproduction of the (quite noisy) energy flow that accompanies hard scatterings in hadron-collider events.

It should be stressed that these multiple parton interactions are a separate effect from the multiple proton interactions that may occur in each collision event in the LHC. These multiple proton collisions are referred to as pileup, and are not included in the definition of the underlying event.

No precise model exists to reproduce the underlying event activity. This activity is instead also adjusted to reproduce available experimental data. A specific set of chosen parameters for a generator is referred to as a “tune”.

The two Monte Carlo generators used in this analysis are summarized below, indicating the particular versions and tunes that were implemented.

Pythia

PYTHIA event generator has been used extensively for e^+e^- , ep , $pp/p\bar{p}$ at LEP, HERA, and Tevatron, and during the last 20 years has probably been the most used generators for LHC physics studies. PYTHIA contains an extensive list of hardcoded subprocesses, over 200, that can be switched on individually. These are mainly $2 \rightarrow 1$ and $2 \rightarrow 2$, some $2 \rightarrow 3$, but no multiplicities higher than that. Consecutive resonance decays may of course lead to more final-state particles, as will parton showers.

As mentioned above, in this MC generator, showers are ordered in trans-

verse momentum [26] both for ISR and for FSR. Also MPIs are ordered in p_T [27]. Hadronization is based solely on the Lund string fragmentation framework.

For the results presented in this thesis simulated samples of dijet events from proton-proton collision processes were generated with PYTHIA 6.423 [23]. The ATLAS AMBT2 tune of the soft model parameters was used [28]. This tune attempts to reproduce the ATLAS minimum bias charged particle multiplicity and angular distribution measurements and the ATLAS measurements of charge particle and p_T density observed collinear and transverse to the high-energy activity.

For systematic comparisons, a set of additional tunes, called the Perugia tunes [29] were also used. These tunes utilize the minimum bias and p_T density measurements of CDF to model the underlying event, hadronic Z^0 decays from LEP to model the hadronization and final state radiation, and Drell Yann measurements from CDF and $D0$ to model the initial state radiation. In particular, the Perugia 2011, which is a retune of Perugia 2010 [30] includes 7 TeV data from 2011 data taking.

Herwig++

HERWIG++ [24] is based on the event generator HERWIG (Hadron Emission Reactions With Interfering Gluons), which was first published in 1986 and was developed throughout the LEP era. HERWIG was written in Fortran, and the new generator, Herwig++ developed in C++. Some distinctive features of Herwig++ are: Angular ordered parton showers and cluster hadronization, and hard and soft multiple partonic interactions to model the underlying event and soft inclusive interactions [?].

This MC generator was used for systematic uncertainties studies. The

version utilized was version 2.4.2 released in 2009.

In order to use events produced by Monte Carlo generators to model events that one might observe with the detector, the output of these generators is passed through a detector simulation model. ATLAS uses the GEANT4 [31] toolkit to simulate the passage of particles through the detector material. This includes models for the production of additional particles caused by inelastic scattering off of electrons and nuclei, as well as ionization and absorption by active detector elements.

2.3.2 Jet algorithms

As described above, quarks and gluons cannot be directly observed. Quarks and gluons hadronise, leading to a collimated spray of energetic hadrons, a jet. By measuring the jet energy and direction one can get close to the idea of the original parton. But one parton may form multiple experimentally observed jets, for example due to a hard gluon emission plus soft and collinear showering. Then, in comparing data to theory and MC programs predictions a set of rules for how to group particles into jets is needed. A jet algorithm, together with a set of parameters and a recombination scheme (how to assign a momentum to the combination of two particles) form a jet definition.

By using a jet definition a computer can take a list of particle momenta for an event (be they quarks and gluons, or hadrons, or calorimeter depositions), and return a list of jets. One important point to remark is that the result of applying a jet definition should be insensitive to the most common effects of showering and hadronization, namely soft and collinear emissions. This is illustrated in Fig. 2.1.

Traditionally, jet algorithms have been classified into two categories: cone and sequential recombination algorithms.

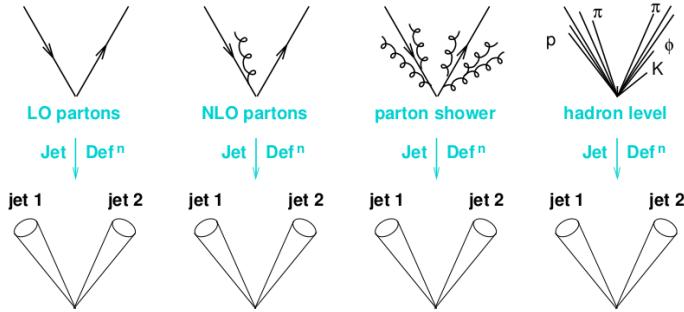


Figure 2.1: The application of a jet definition to a variety of events that differ just through soft/collinear branching and hadronization should give identical jets in all cases [32].

Cone-like algorithms are based on the collinear nature of gluon radiation and the parton shower described above. The decay products of quarks and gluons and their emissions will tend to form a cone of particles in the $\eta - \phi$ plane⁵ as they propagate. A cone algorithm will work as follows⁶: first, it sorts all particles in the event according to their momentum, and identifies the one with largest p_T . This is referred to as a seed particle. Then a cone of radius R in $\eta - \phi$ is drawn around the seed. The direction of the sum of the momenta of those particles is identified and if it doesn't coincide with the seed direction then the sum is used as a new seed direction, and it iterates until the sum of the cone contents coincides with the previous seed (this type of algorithm is “iterative” cone since it iterates the cone direction). This is how

⁵The azimuthal angle ϕ is measured around the beam axis, and the polar angle θ is the angle from the beam axis. The pseudorapidity is defined as $\eta = \ln \tan(\frac{\theta}{2})$. The transverse momentum p_T is defined in the plane transverse to the beam motion. The distance ΔR in the pseudorapidity-azimuthal angle space is defined as $\Delta R = \sqrt{\Delta\eta^2 + \Delta\phi^2}$. See section 3.2 for more details on the ATLAS Coordinate system.

⁶This is how CMS cone algorithm, used for the preparation for the LHC running, works.

a stable cone is reached. A difficulty and major drawback in this procedure is the use of the transverse momentum of the particle to select the first seed. This definition is collinear unsafe, i.e. a splitting of the hardest particle into a nearly collinear pair can have the consequence that another, less hard particle, pointing in a different direction suddenly becomes the hardest in the event, leading to a different final set of jets. There are many other variants of cone algorithms, and nearly all suffer from problems of either collinear safety, or infrared safety (an extra soft particle creates a new seed, which can lead to an extra stable cone being found). A fix for these problems came in a algorithm called Seedless Infrared Safe Cone (SISCCone) [33].

Recombination algorithms, on the other hand, are both collinear and infrared safe. And for this reason, they can be used in calculations to any order in perturbation theory. The term recombination is appropriate given that these algorithms work as if they were inverting the sequence of splittings of the parton shower. In general, recombination algorithms operate by successively combining pairs of particles using a distance metric, d_{ij} . At hadron colliders, due to the fact that one of the incoming partons may continue along the beam, for every pair of particles this metric is compared to a so-called “beam distance”, d_{iB} , and only when $d_{ij} < d_{iB}$ the particle pair is combined and considered for subsequent clustering steps.

ATLAS (and also CMS) has chosen the anti- k_t [34] algorithm as the default jet algorithm for use in physics analysis. This recombination algorithm as well as the Cambridge-Achen algorithm [?] (C/A) are extensions of the original k_t algorithm developed for the analysis of multi-jet events at e^+e^- colliders [35] and subsequently extended for use at hadron colliders [36, 37]. In this thesis, the k_t algorithm was used for jet substructure studies, see section 2.3.3.

The original k_t algorithm implements the following (2.1) distance metric between particles i and j

$$d_{ij} = \frac{2E_i E_j (1 - \cos \theta_{ij})}{Q^2}, \quad (2.1)$$

where Q is the total energy in the event, E_i is the energy of particle i and θ_{ij} the angle between particles i and j . In the collinear limit, d_{ij} is related to the relative transverse momentum between particles i and j (hence the name k_t algorithm), normalized to the total visible energy. The particles are combined if the minimum d_{ij} , d_{min} , is below a certain threshold, y_{cut} . The jet multiplicity depends on the value of y_{cut} , as a lower value will result in more soft or collinear emissions surviving as jets. As mentioned above, for hadron colliders, the notion of a beam distance is added. A distance scale, $\Delta R = \sqrt{\Delta y^2 + \Delta \phi^2}$, is introduced to define the typical radius for a jet, effectively replacing y_{cut} . In this case for every pair of particles a new distance is defined, (2.2),

$$d_{ij} = \min(p_{ti}^2, p_{tj}^2) \frac{\Delta R_{ij}^2}{R^2} \quad (2.2)$$

and the beam distance, $d_{iB} = p_{ti}^2$. The algorithm proceeds by searching for the smallest of the d_{ij} and the d_{iB} . If it is a d_{ij} then particles i and j are recombined into a single new particles. If it is a d_{iB} then i is removed from the list of particles, and called a jet. This is repeated until no particles remain.

As opposed to cone algorithms, for the k_t algorithm, the jets have quite irregular shapes, and particles with $\Delta R_{ij} > R$ can still be clustered within the jet. This is a problem when, for example, an irregularly shaped jet happens to extend into poorly instrumented detector regions. Another drawback of this definition is that soft particles are clustered first. This has the po-

tential to introduce complications when the detector noise of energy density fluctuations are large.

A feature of the k_t algorithm that is attractive is that it does not produce jets but it also assigns a clustering sequence to the particles within the jet. It is possible then to undo the clustering and to look back at the shower development history. This has been exploited in a range of QCD studies, and also in searches of hadronic decays of boosted massive particles and it will be used here for the search of two-pronged jets in gluon splitting.

The prescription above may be generalized beyond the k_t algorithm. By inverting the power law in the particle distance metric, d_{ij} , the anti- k_t algorithm is obtained. The particle distance metric used by this algorithm is,

$$d_{ij} = \min(p_{ti}^{-2}, p_{tj}^{-2}) \frac{\Delta R_{ij}^2}{R^2} \quad (2.3)$$

and the beam distance, $d_{iB} = p_{ti}^{-2}$. This definition results in the clustering of the hardest emissions first. This has several benefits in the context of high-luminosity hadron collisions.

Note that the anti- k_t algorithm does not provide useful information on jet substructure if a jet contains two hard cores, then the k_t (or C/A) algorithms first reconstruct those hard cores and merge the resulting two subjets. The anti- k_t will often first cluster the harder of the two cores and then gradually agglomerate the contents of the second hard core.

These algorithms, and more, are implemented in FASTJET [38] software package for jet-finding.

2.3.3 Jet substructure

The first evidence of jet structure resulted from the study of the spacial distribution and multiplicity of particles in the event phase space in hadron

production in e^+e^- collisions [20]. Generally, all final hadronic states in both in $pp/p\bar{p}/e^+e^-$ collisions can be explored in terms of the structure and shape of the event energy flow by means of the so called “event shape” variables. This family of variables attempts to extract information about the global geometry of an event, usually distinguishing between di-jet events and multijet final states. Such variables have been successfully utilized in many SM measurements and BSM searches, see for example [39][40].

Although very useful, event shape variables are not sensitive to the detailed structure and distribution of energy inside a particular jet. In SM and new physics searches, tools for the identification of individual objects that might be signature of new particles are desired. At the LHC, many of the particles considered to be heavy at previous accelerators will be frequently produced with a transverse momentum greatly exceeding their rest mass, like the electro-weak gauge bosons W^\pm and Z , the top quark, the Higgs boson (or bosons) and possibly other new particles in the same mass range. These boosted objects, produced either by recoil against other energetic objects or from decays of even heavier BSM particles, upon decay can give rise to a highly collimated topology too close to be resolved by standard jet algorithms. A method for selecting these jets would allow for the study of their properties. This interest led to the development of a wide range of sophisticated tools in the last years [41, 42] that allow the analysis of the substructure of the ensuing jet and reveal its heavy-particle origin.

Jet substructure methods probe the internal structure of jets from a detailed study of its constituents. These techniques have been first implemented for distinguishing boosted SM hadronic objects from the background of jets initiated by light quarks and gluons, see for example [43], but they have been also successfully used in other applications, including separating quark

jets from gluon jets [44] and identifying boosted decay products in new physics searches [45].

Jet shapes, which are event shape-like observables applied to single jets, are an effective tool to measure the structure of individual jets [46]. The shape of a jet not only depends on the type of parton (quark or gluon) but is also sensitive to non-perturbative fragmentation effects and underlying event contributions [47].

In chapter 5, several distinguishing characteristics between jets originating from single b -quarks and jets containing two close-by b -hadrons are determined using the techniques of jet substructure.

2.4 Heavy flavor jet production

Heavy flavor quarks enter in many collider searches, notably because they are produced in the decays of various SM particles (top quarks, the Z boson and the Higgs boson, if light), and of numerous particles appearing in proposed extensions of the SM. Heavy flavour quark production in hadronic collisions may be subdivided into three classes depending on the number of heavy quarks participating in the hard scattering. The hard scatter is defined as the $2 \rightarrow 2$ subprocess with the largest virtuality (or shortest distance) in the hadron-hadron interaction [48].

- Heavy flavor creation (FCR): two heavy quarks are produced in the hard subprocess. Being Q the heavy flavor quark, at leading order this process is described by $gg \rightarrow Q\bar{Q}$ and $q\bar{q} \rightarrow Q\bar{Q}$.
- Heavy flavor excitation (FEX): the heavy flavour quark excitation can be depicted as an initial state gluon splitting into a heavy quark pair, where one of the heavy quarks subsequently enters the hard subprocess.

- Gluon splitting (GSP): no heavy quarks participate in the hard subprocess in this case, but they are produced in $g \rightarrow Q\bar{Q}$ branchings in the parton shower.

Example of Feynman diagrams for QCD b -quark production up to NLO are shown in Fig. 2.2.

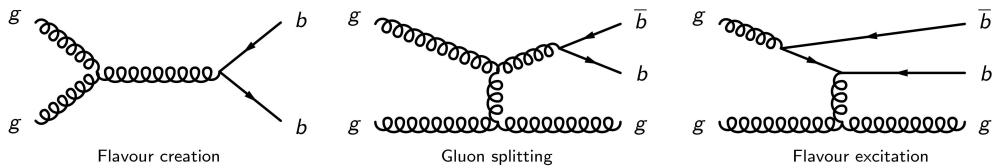


Figure 2.2: Representative diagrams of the three channels contributing to QCD b -quark production up to NLO. The flavour creation channel (left) is the only one present at LO. At NLO, two new channels open up, referred to as gluon splitting (center) and flavour excitation (right).

The definition above is not strict, but can be used as a basis for the understanding of the characteristics of heavy flavour quark production.

The simplest and most fundamental measurement of heavy-quark jet production is the inclusive heavy-quark jet spectrum, which is dominated by pure QCD contributions. Studies of QCD bottom production are important in their own right because of the correspondence between parton level production and the observed hadron level, and their potential to provide information on the b -quark parton distribution function, a component of the proton structure thought to be generated entirely perturbatively from the QCD evolution equations of the other flavours. The theoretical calculation of the inclusive b -jet spectrum presents rather important uncertainties ($\sim 50\%$), considerably larger than those for the light jet inclusive spectrum ($\sim 10 - 20\%$) [1].

A review of the origin of these uncertainties is presented by Banfi, Salam and Zanderighi in reference [49]. They arise from the poor convergence of the perturbative series, as evidenced by a large value of the K -factor, the ratio of the next-to-leading order (NLO) to the leading order (LO) cross section, in the p_T range covered by the LHC. This is illustrated in Fig. 2.3. The observed K values (6 to 10) indicate that the NLO result cannot be an accurate approximation to the full result. It is for this reason that the scale dependence (middle panel in Fig. 2.3) is large.

The fact that the perturbative series is very poorly convergent is related to the different channels for heavy quark production. While at LO only the FCR channel is present, at NLO the FEX and GSP channels open up⁷. In the gluon splitting process, one of the final-state light partons (at NLO always a gluon) splits collinearly into a $b\bar{b}$ pair that a clustering algorithm can classify within the same jet. A jet containing both b and \bar{b} is considered to be just a b -jet in standard definitions.

The various channels can be approximately separated in a parton shower Monte Carlo generator such as HERWIG or PYTHIA. These MC generators include NLO effects, and one can determine the underlying hard process from the event record. Their relative contributions to the total b -jet spectrum are shown in the bottom panel of Fig. 2.3. It is found that the LO channel has a much smaller contribution than the FEX and the GSP channels, which receive strong enhancement from collinear logarithms, going as $\alpha_s^2(\alpha_s \ln(p_T/m_b))^n$ for flavour excitation [19] and $\alpha_s^2 \cdot \alpha_s^n \ln^{2n-1}(p_T/m_b)$ for gluon splitting ($n \geq 1$) [51].

⁷It is sometimes stated that it makes no sense, beyond LO, to separately discuss the different channels, for example because diagrams for separate channels interfere. However, each channel is associated with a different structure of logarithmic enhancements, $\ln^n(p_T/m_b)$, and so there is distinct physical meaning associated with each channel.

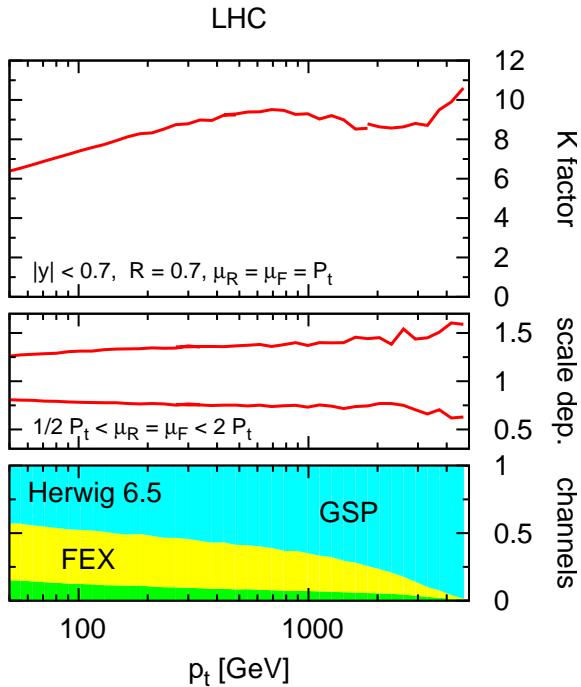


Figure 2.3: Top: K -factor for inclusive b -jet spectrum taken from [49], clustering particles into jets using the k_t jet-algorithm [37] with $R=0.7$, and selecting jets in the central rapidity region ($|y| < 0.7$). Middle: scale dependence obtained by simultaneously varying the renormalisation and factorisation scales by a factor two around p_T , the transverse momentum of the hardest jet in the event. Bottom: breakdown of the Herwig [50] inclusive b -jet spectrum into the three major underlying channels, flavor creation (FCR) flavor excitation (FEX) and gluon splitting (GSP).

Ref. [49] proposes a new observable to free the heavy-flavour spectrum calculation from collinear logarithms, and improve the accuracy of the theoretical prediction, by not including in the production cross-section the contribution from double b jets. Final-state logarithms are removed by employing a recently developed jet reconstruction scheme, the flavour- k_t algorithm [52], which maintains the correspondence between partonic flavour and jet flavour. Specifically, jets containing a b -quark and a b -antiquark, which in a parton shower MC generator are produced $\sim 95\%$ of the time by the GSP channel, are labeled in an IR-safe way as light jets and removed from the b -jet spectrum. The initial-state (FEX) collinear logarithms can be resummed by using a b -quark parton distribution functions. With this algorithm the K -factor for the differential heavy-jet spectrum cross-section is shown not to exceed a value of $K = 1.4$, with a factor of four reduction in the theoretical (scale variation) uncertainties.

Succesfully identifying jets with two b -hadrons, the products of the b -quark or b -antiquark hadronization, can also provide an important handle to understand, estimate and/or reject b -tagged backgrounds to SM and new physics searches at the LHC.

SM physics analyses that rely on the presence of single b -jets in the final state, such as top quark physics, either in the $t\bar{t}$ or the single top channels, and associated Higgs production: $WH \rightarrow \ell\nu b\bar{b}$ and $ZH \rightarrow \nu\nu b\bar{b}$, suffer from the reducible background from QCD, which can produce double b -hadron jets as discussed above, and the irreducible background due to W bosons produced in association with b -quarks. Figure 2.4 shows the two diagramas for $W+b$ production. While at LO only single b -jets are present, at NLO jets containing two b -hadrons are expected due to the contribution of a diagram containing a $g b\bar{b}$ vertex. The b -quark pair is produced at small angles and

can be often reconstructed as one merged jet.

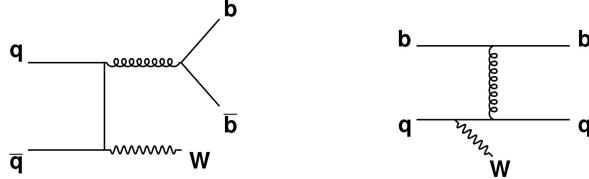


Figure 2.4: Feynman diagrams for W production in association with b quarks.

The relevance of double b -hadron jets is supported by NLO calculations of the production of W bosons and two jets with at least one b quark at the LHC for jet $p_T > 25$ GeV, and $|\eta| < 2.5$ [53] indicate that the cross section for $W(b\bar{b})j$ is almost a factor of two higher than $Wb\bar{b}$, and about a third of Wbj , where $W(b\bar{b})j$ denotes the case in which the two b quarks are merged into the same jet.

Jets containing a single b -quark or antiquark also enter in many BSM collider searches, notably because b -quarks are produced in the decays both of heavy SM particles (top quarks, the Z boson and the Higgs boson), and of particles appearing in proposed extensions of the SM. An example is the search for supersymmetry in the framework of generic R -parity conserving models [54]. The superpartners of quarks and gluons could be copiously produced via the strong interaction at the LHC. The partners of the right- and left-handed quarks, \tilde{q}_L and \tilde{q}_R , can mix to form two mass eigenstates and, since mixing is proportional to the corresponding fermion masses, it becomes more important for the third generation producing sbottom and stop significantly lighter than the other squarks. In this model, thus, sbottom and stop production is expected to dominate. As they chain decay to b -quarks and the lightest supersymmetric particle, the signature for this channel is missing transverse energy plus (single) b -jets. The ability to distinguish single b -jets

from jets containing two b -hadrons is thus here of wide application to reduce SM backgrounds giving rise to close-by $b\bar{b}$ pairs.

The study of $b\bar{b}$ jets from gluon splitting is an ideal testbed for exploring jet substructure in data, as it provides a large supply of boosted, merged jets. Furthermore, understanding $g \rightarrow b\bar{b}$ jets is important as they are themselves the background to boosted object searches, like $Z \rightarrow b\bar{b}$ or $H \rightarrow b\bar{b}$. Understanding the much more common QCD events with double b -hadron jets will be essential before attempting to measure more rare final states.

Chapter 3

The ATLAS detector at the LHC

3.1 The Large Hadron Collider

The Large Hadron Collider (LHC) [55] is a proton-proton (pp) synchrotron located in the previous Large Electron Positron (LEP) collider tunnel at CERN Laboratory, just outside the city of Geneva (Switzerland), approximately 100 m underground. It is designed to collide bunches of up to $\sim 10^{11}$ protons every 25 ns at a center-of-mass energy of 14 TeV (seven times the 2 TeV reached by the Tevatron accelerator at Fermilab Laboratory, in Chicago).

The experiments analyzing the collisions produced by the LHC are distributed around the 27 km ring at the various interaction points. The ATLAS experiment is located at Point 1, which is closest to the main CERN site. Point 5 houses the other general purpose detector, CMS. ALICE and LHCb experiments are located at Point 2 and Point 8, respectively. The former is designed to investigate heavy ion collisions; the latter, to investigate rare decays of b -mesons. The layout of these four experiments along the LHC ring is shown in Fig. 3.1.

Proton beams are formed, before insertion into the main LHC ring, using

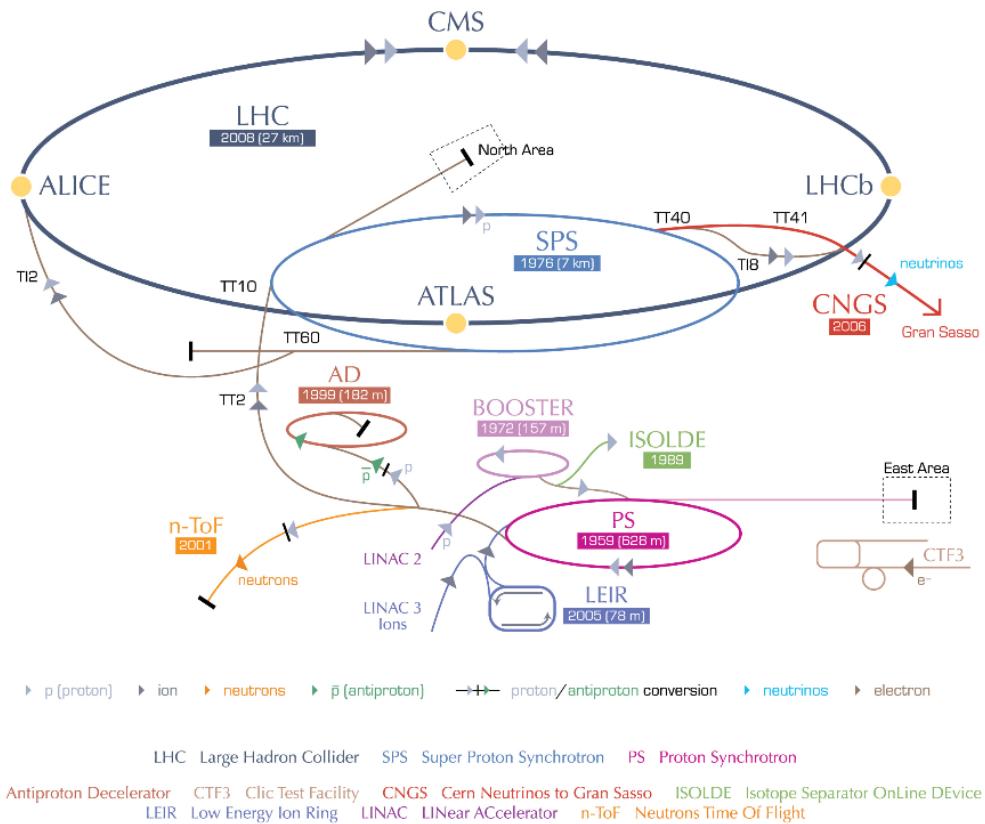


Figure 3.1: The CERN accelerator complex, showing the injection system, along with each component's date of construction, and the placement of the four main experiments.

a succession of smaller machines with increasingly higher energies, as shown in Fig. 3.1. The chain begins as protons are injected into the PS Booster (PSB) at an energy of 50 MeV from Linac2. The booster accelerates them to 1.4 GeV. The beam is then fed to the Proton Synchrotron (PS) where it is accelerated to 25 GeV. At design strength, the bunch structure, known as a bunch train, contains 72 bunches of protons upon entry to the Super Proton Synchrotron (SPS). The SPS accumulates up to four fills of 72 bunches from the PS and accelerates them to 450 GeV, with a bunch spacing of ~ 25 ns. They are finally transferred to the LHC (both in a clockwise and an anticlockwise direction) where they are accelerated for 20 minutes to their nominal energy of 7 TeV. Beams will circulate for many hours inside the LHC beam pipes under normal operating conditions.

The bunch structure is a direct consequence of a radio frequency (RF) acceleration scheme used to attain the desired high proton beam energy. In RF acceleration, particles travel through a series of time-varying electrical fields and they can only be accelerated when the RF field has the correct orientation when particles pass through an accelerating cavity, which happens at well specified moments during an RF cycle. The result of a sequence of RF accelerations is several bunches of protons. It is important to note that when we speak about “beams” we refer to many bunches of protons separated by some uniform distance. Increasing the number of bunches is one of the ways to increase luminosity in a machine (more about luminosity in subsection 3.1.1). At designed beam intensity, when the bunches cross, there will be a maximum of about 20 collisions.

A large magnetic field is needed to guide and maintain the beam particles in their circular orbit. The needed field is achieved using superconducting electromagnets built from NbTi coils that operate in a superconducting

state, efficiently conducting electricity without resistance or loss of energy. The currents through the coils produce magnetic fields perpendicular to the direction of motion of the protons that deflect the protons into their orbits. The whole magnetic system comprises 1232 dipole magnets of 15 m length which are used to bend the beams, and 392 quadrupole magnets, each 5-7 m long, to focus the beams. At a peak beam energy of 7 TeV, the dipoles need to produce an 8.33 T magnetic field, requiring a current of ~ 12 kA. In order to deliver the current densities and magnetic field required for 7 TeV proton beams, the magnets are kept at 1.9 K by circulating superfluid helium.

The first pp collisions produced by the LHC occurred on November 23 2009, at the SPS extraction energy of 450 GeV per beam. Very quickly after, on December 8, ATLAS and CMS detectors started recording data at energy of 2.36 TeV. By this time the LHC became the highest energy accelerator in the world. During this period, bunch intensities were limited by machine-protection considerations to 1.5×10^{10} protons.

In February 2010, the LHC was commissioned once more with 450 GeV beams, and a series of tests were performed to ensure that the magnet systems could operate safely at the currents necessary to control 3.5 TeV beams. This was followed by the very first collisions at 7 TeV center-of-mass energy on March 30. During the 2010 run the beam parameters were tuned (the beam widths squeezed and the number of protons per bunch and the number of bunches in each beam increased) in order to increase the beam intensity. In particular, as the intensity of the beams increased, the mean number of interactions per bunch crossing increased.

The data samples analyzed in this thesis correspond to proton-proton collisions at $\sqrt{s} = 7$ TeV delivered by the LHC and recorded by ATLAS between May and November 2011, with the LHC running with 50 ns bunch

spacing. Table 3.1 summarizes the basic beam parameters expected for design energy and luminosity and the beam parameters as of May 2011. The LHC performance steadily improved during 2011. The average number of interactions per bunch crossing throughout the data-taking period considered rapidly increased approximately from ~ 3 to 8 until (northern hemisphere) summer 2011, with a global average for this period of ≈ 6 . Starting in August 2011 and lasting through the end of the proton run, this number ranged from approximately 5 to 17, with an average of about 12. This evolution is illustrated in Fig. 3.2, which shows the maximum mean number of collisions per beam crossing versus day in 2011.

Parameter	2011 runs	Design
Center-of-mass energy [TeV]	7	14
Instantaneous luminosity [$\text{cm}^{-2}\text{s}^{-1}$]	3.65×10^{33} (year peak)	10^{34}
Bunches per beam	38 (May)	2808
Protons per bunch	0.8×10^{11} (May)	1.5×10^{11}
Mean interactions per crossing	6 to 12 (year average)	23

Table 3.1: Summary of beam conditions during the 2011 7 TeV runs and those foreseen at design energy and luminosity.

3.1.1 Luminosity and pile-up

The rate of events produced by the colliding beams depends on the luminosity of the collisions, which is a measure of the number of events per second per unit cross section, typically measured in units cm^2s^{-1} . The number of events of a particular process, then, is given by the product of the integrated luminosity, $\int dt L$, and the cross section of the process, σ_{event} . The

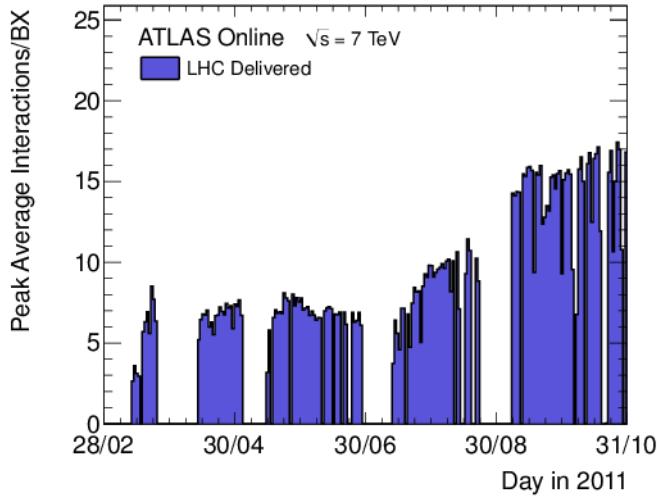


Figure 3.2: The maximum mean number of events per beam crossing versus day in 2011.

integrated luminosities are typically quoted in units of inverse picobarns, $\text{pb}^{-1} = 10^{-36}\text{cm}^2$. In order to measure processes with very little cross sections a very high luminosity is required.

The delivered luminosity can be written as [56]:

$$L = \frac{n_b f_r n_1 n_2}{2\pi \Sigma_x \Sigma_y} \quad (3.1)$$

where n_b is the number of colliding bunch pairs, n_1 and n_2 are the bunch populations (protons per bunch) in beam 1 and beam 2 respectively (together forming the bunch charged product), f_r is the machine revolution frequency, and Σ_x and Σ_y are the width and the height of the proton beams.

The number of protons per bunch, the number of bunches per beam, and the revolution frequency are all set by the beam operators. The widths of the proton beams are measured in a process known as a Van der Meer (*vdm*) scan [57]. In a *vdm* scan, the beams are separated by steps of a known

distance. The collision rate is measured as a function of this separation, and the width of a gaussian fit to the distributions yields the width of the beams in the direction of the separation.

The total integrated luminosities provided by the LHC and recorded by ATLAS in 2011 are shown in Figure 3.3. These events form the dataset analyzed in this thesis. By means of the beam-separation or vM scans, as well as other techniques to measure the bunch charged product, the ATLAS Collaboration has determined that the uncertainty on its luminosity measurement is $\delta L = \pm 3.7\%$. For a complete description of the methods used and the systematic errors evaluated see reference [56].

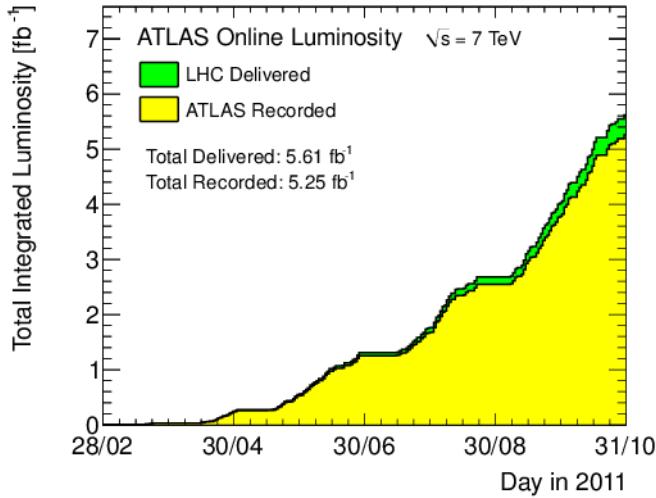


Figure 3.3: Total luminosity delivered by the LHC and recorded by ATLAS during the 2011 $\sqrt{s} = 7$ TeV proton-proton run

As anticipated, due to the cross-section for interaction and the number of protons per bunch, the possibility to observe multiple pp interactions per bunch crossing increases proportionally. This phenomenon, referred to as “pile-up”, can really occur in two distinct forms. The first form is the pres-

ence of multiple pp collisions (different from the interaction of interest) in the same bunch crossing, referred to as “in-time” pile-up. The second form of pile-up takes place due to electronic integration times within the detector. Certain detector components are actually sensitive to multiple bunch crossings due to the long electronic signals generated in the response to energy depositions or charge collection. One or more pp collisions in a bunch-crossing different from that which produced the collision of interest can then affect the measurement. This form of pile-up is referred to as “out-of-time” pile-up and will become more and more important as the LHC bunch spacing gets closer to the nominal value, 25 ns.

The fraction of events with pile-up increased significantly since the data taking started. The experimental signature of this fact is obtain via the number of reconstructed primary vertices, or NPV. The effect of the event NPV is an important concern for the measurement of jet properties and will be discussed in the next chapters.

3.2 The ATLAS Detector

The ATLAS detector [58] is one of the two general purpose particle detectors built for probing pp collisions at the LHC. As it was described in the previous section, inside the LHC, bunches of up to 10^{11} protons will collide 40 million times per second to provide 14 TeV proton-proton collisions at a nominal luminosity of $10^{34}\text{cm}^{-2}\text{s}^{-1}$. These high interaction rates and energies, as well as the requirements for high precision physics measurements set the standars for the design of the detector. At even 7 TeV center-of-mass energy, the LHC interactions result in high particle multiplicity, requiring fine

detector granularity, and particle production at forward rapidity, requiring large detector angular coverage.

To achieve these performance goals, a design consisting of multiple detector sub-systems with cylindrical symmetry around the incoming beams is used as shown in Fig. 3.4. Closest to the interaction point the inner tracking detector is placed, providing charged particle reconstruction. The magnet configuration comprises a thin superconducting solenoid surrounding the inner detector cavity, and three large superconducting toroids (one barrel and two end-caps) arranged with an eight-fold azimuthal symmetry around the calorimeters. This fundamental choice has driven the design and size (44 m in length and 25 m in height) of the rest of the detector. Outside the solenoid, a calorimeter system performs electron, photon, tau, and jet energy measurements¹. Finally, the calorimeter is surrounded by the muon spectrometer where an array of muon drift chambers perform muon identification and momentum measurements.

The ATLAS detector coordinate system is used to describe the position of particles as they traverse these subdetectors. It is a right-handed coordinate system, with z pointing along the beam direction, positive x pointing toward the center of the LHC ring, and positive y pointing up. The $x - y$ plane is referred to as the transverse plane, and the z direction as the longitudinal direction. The azimuthal angle ϕ is measured as usual around the beam axis, and the polar angle θ is the angle from the beam axis. The pseudorapidity is defined as $\eta = \ln(\tan(\frac{\theta}{2}))$, regions of low η are referred to as “central”, and regions of high η are referred to as “forward”. The transverse momentum p_T is defined in the $x - y$ plane unless stated otherwise. The distance ΔR in the pseudorapidity-azimuthal angle space is defined as $\Delta R = \sqrt{\Delta\eta^2 + \Delta\phi^2}$.

¹

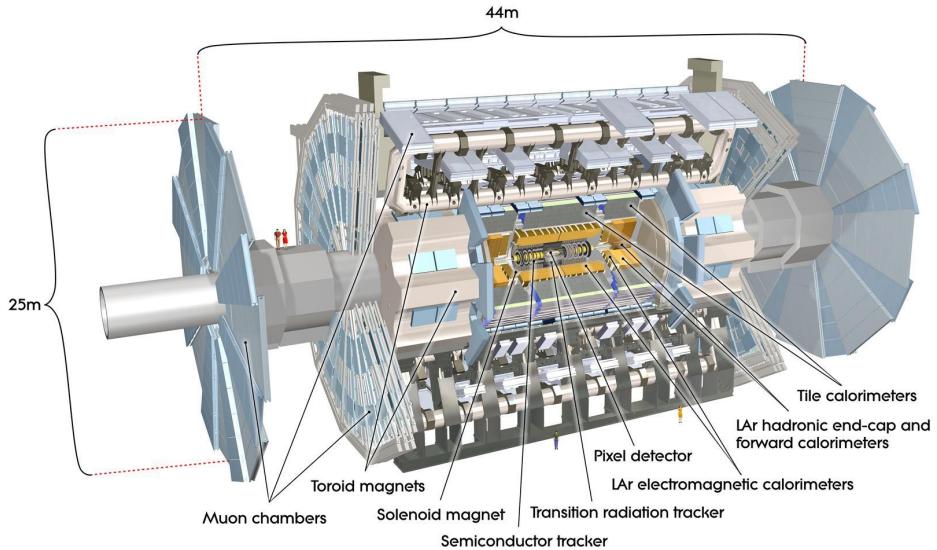


Figure 3.4: El detector de ATLAS

To meet the extremely high demands that the LHC luminosity places on the speed with which ATLAS must record data, a dedicated trigger and data acquisition (TDAQ) system is used. The interaction rate at the design luminosity is approximately 1 GHz, while the event data recording, based on technology and resource limitations, is limited to about \sim 200 Hz. This requires a high rejection of minimum-bias processes while maintaining maximum efficiency for the new physics. The Level-1 (L1) trigger system uses a subset of the total detector information to make a decision on whether or not to continue processing an event, reducing the data rate to approximately 75 kHz (limited by the bandwidth of the readout system, which is upgradeable to 100 kHz). The subsequent two levels, collectively known as the high-level trigger (HLT), are the Level-2 (L2) trigger and the Event Filter (EF). They provide the reduction to a final data-taking rate of approximately 200 Hz.

3.2.1 Inner tracking system

The inner tracking system or Inner Detector (ID) is composed of three sub-detectors: the pixel detector, the semiconductor tracker (SCT) and the transition radiation tracker (TRT). The goal of these three is to provide charged particle trajectory reconstruction and momentum measurements with an overall acceptance in pseudorapidity of $|\eta| < 2.5$ and full ϕ coverage.

The sensors which built this system register signals, referred to as “hits”, in response to the passage of charged particles. The ID is immersed in a 2 T magnetic field, generated by the central solenoid. The positions of the registered hits are combined to form tracks, with the radius of curvature of the tracks (caused by the presence of the magnetic field) providing a measurement of the particle’s transverse momentum. The track reconstruction efficiency ranges from 78% at $p_T^{track} = 500$ MeV to more than 85% above 10 GeV, averaged across the full η coverage [59]. The transverse momentum resolution of $\sigma_{p_T}/p_T = 0.05$ [60] (upper bound) and a transverse impact parameter resolution of $\sim 20 \mu\text{m}$ for high momentum resolution particles in the central η region[61] are primarily achieved through the use of high precision subsystems within the ID.

The pixel detector, SCT, and TRT sensors are arranged on concentric cylinders around the beam axis, known as barrel layers, and on disks perpendicular to the beam at either end of the barrel, known as end-caps. A more complete description of these systems is given below. The overall layout of the inner detector is shown in Fig. 3.5.

The Pixel detector

The pixel detector consists of three concentric barrel layers. The innermost one, the so called “*b*-layer” due to its role in identifying *b*-quarks initiated

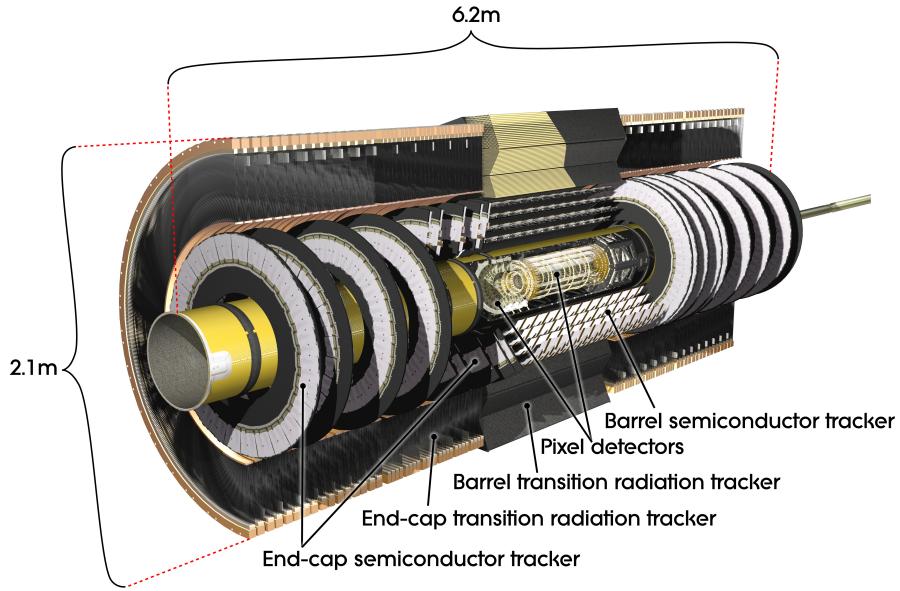


Figure 3.5: Layout of the ATLAS Inner Detector.

jets, is located at 5 cm from the interaction region. Three additional disks are located at each end-cap, producing typically three pixel position measurements per charged particle track. Each layer or disk is instrumented with modules that form the basic unit of data acquisition, each with 47,232 pixels. All pixel sensors are identical and have a minimum pixel size in $r - \phi \times z$ of $50 \times 400 \mu m^2$. The intrinsic accuracies in the barrel are $10 \mu m$ in $r - \phi$ and $115 \mu m$ along z , or along r in the end-caps. The pixel detector has approximately 80.4 million readout channels, an order of magnitude more readout channels than the rest of ATLAS combined, and it extends to a total length of $z \sim \pm 650$ mm and radius of $r \sim 150$ mm, providing good reconstruction efficiency for tracks up to $|\eta| < 2.5$.

The SCT

The SCT consists of four barrel layers and nine end-cap layers surrounding the pixel detector, resulting in at least four hits along every charged particle track. The SCT barrel reaches to $z \sim \pm 750$ mm and $r \sim 515$ mm, while the end-cap covers out to $z \sim \pm 2720$ mm and $r = 560$ mm. There are 15,912 SCT module sensors, each 12.8 cm long and approximately 285 μm thick.

In the barrel region, these modules use small-angle (40 mrad) stereo strips to measure both coordinates, with one set of strips in each layer parallel to the beam direction, measuring the ϕ coordinate directly . In the end-cap region, the detectors have a set of strips running radially and a set of stereo strips at an angle of 40 mrad. The mean pitch of the strips is 80 μm . The intrinsic accuracies per module in the barrel are 17 μm in $r - \phi$ and 580 μm in z (or r in the end-caps). The total number of readout channels in the SCT is approximately 6.3 million. A hit is registered only if the pulse height in a channel exceeds a preset threshold (~ 1 fC). The charge measured in the strip is then recorded into a memory buffer that is only read out and used for tracking if a trigger is received signaling that the event should be considered in more detail.

The TRT

The TRT surrounds the silicon detectors and is comprised of up to 76 layers of longitudinal straw tubes in the barrel, extending to $z \sim \pm 710$ mm and $r \sim 1060$ mm, and 160 radial straw planes in each end-cap cylinders, reaching $z \sim \pm 2710$ mm and $r \sim 1000$ mm.

The TRT sensors are thin drift tubes consisting of cathode metal straws filled with an ionizing gas mixture of xenon, oxygen, and CO₂, with an anode wire running down the center of the straw. The passage of a charged particle

through the gas produces positive ions and free electrons, which travel to the cathode and anode, respectively, under the influence of an applied voltage of 1600 V. Comparing the time that the signals are received at the cathode and the anode gives a drift time measurement that can be used to calculate the impact parameter of the particle. This method gives no information on the position along the length of the straw.

To give the best resolution of particle trajectories as they bend in the solenoidal field, the straws lie along the beam direction in the barrel and radially in the end-caps. The straw diameter of 4 mm causes a maximum drift time of approximately 48 ns and an intrinsic accuracy of 130 μm along the radius of the straw.

In addition to directly detecting charged particles produced by the collision, the TRT also measures the transition radiation induced by the passage of these particles through polypropylene sheets placed between the drift tube straws. Transition radiation refers to the photons emitted by charged particles as they pass from one material into another with a different dielectric constant. These photons yield a much larger signal amplitude than the charged particles, so separate thresholds in the electronics can be used to distinguish the two.

One of the most important tasks of the inner detector is to provide accurate collision vertex identification, exploiting the excellent position resolution and tracking efficiency. Vertices are reconstructed by matching inner detector tracks with $p_T > 150$ MeV back to a common origin.

3.2.2 The Calorimeter System

The purpose of the ATLAS calorimeter system is to measure the energy of electrons, photons, taus and jets, within the pseudorapidity region of $|\eta| < 4.9$

and with full ϕ symmetry and coverage around the beam axis. It also provides fast position and energy measurements to serve as trigger signals for these objects as well as the missing transverse energy.

The calorimeter detector consist of electromagnetic (EM) calorimeter and hadronic calorimeter components. The EM calorimeter provides fine granularity measurements of electrons and photons. Each calorimeter is segmented both transverse to the particle direction, to give position information, and along the particle direction, to chart the development of the particle shower. This permits detailed mapping of EM and hadronic showers in the calorimeter, allowing for studies of the internal structure of hadronic jets and partially giving rise to the high resolution measurements of their energy.

The EM and hadronic calorimeters are sampling calorimeters meaning that they utilize alternating layers of absorber material, composed of heavy atoms that interact with energetic particles and cause them to loose energy, and an active material, that produce a signal in response to the deposited energy.

The calorimeters closest to the beam-line are housed in three cryostats, one barrel and two end-caps. The barrel cryostat contains the electromagnetic barrel calorimeter, and the two end-cap cryostats each contain an electromagnetic end-cap calorimeter (EMEC), a hadronic end-cap calorimeter (HEC), located behind the EMEC, and a forward calorimeter (FCal) to cover the region closest to the beam. These calorimeters use liquid argon as the active detector medium and need to be maintained at a constant temperature of $\sim 88\text{K}$. Liquid argon (LAr) has been chosen for its intrinsic linear behaviour (production of ionization charge as a function of incident charge), its stability of response over time and its intrinsic radiation-hardness.

An illustration of all these components can be found in Fig. 3.6. Further

specifications are given in the next sections.

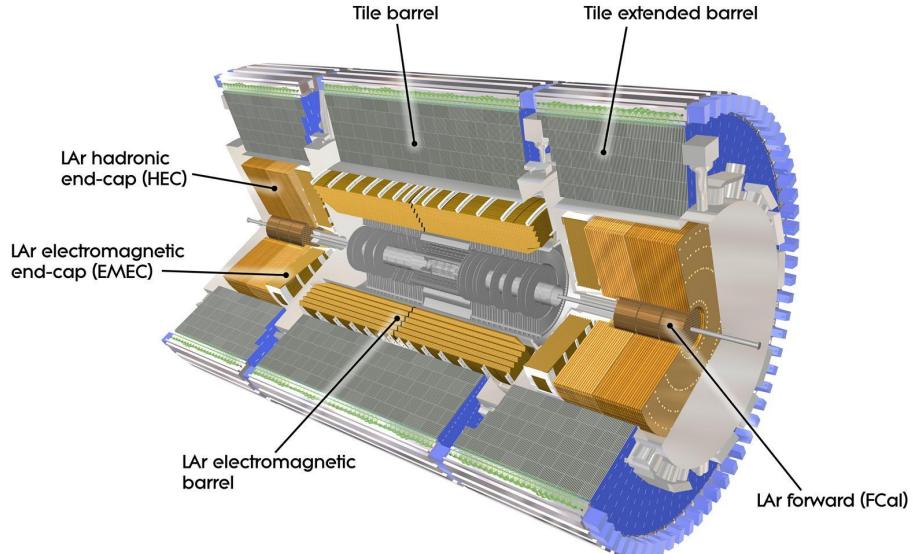


Figure 3.6: Layout of the ATLAS electromagnetic and hadronic calorimeter systems. The total length is ~ 12 m, extending to a maximum radius of 4.25 m.

Liquid argon EM calorimeter

The EM calorimeter uses lead as the absorber and liquid Argon as the active material. A photon traversing the absorber will interact with the heavy nucleus via Compton scattering or the photo-electric effect, producing low-energy electrons, or pair production, producing electron/positron pairs. An electron or positron, in turn, can produce bremsstrahlung photons as it is deflected by the nuclei or produce more charged particles via ionization. Thus each incident photon, electron, or positron produces a shower of photons, electrons, and positrons that lose their energy through successive interactions

in the absorber. The produced particles ionize the liquid argon, and the charge is collected by electrodes located in the liquid argon gap. These electrodes consist of three layers of copper sheets, the outer two kept at high-voltage potential and the inner one used to readout the signal.

To provide full coverage in ϕ without any cracks, an accordion-shaped absorber and electrode geometry is used, shown in Fig. 3.7. This design was chosen to ensure high azimuthal uniformity, a regular liquid argon ionization gap, and a constant sampling fraction within a given detector region. The figure highlights how this geometry is divided among rectangular cells in $\eta \times \phi$ space, the individual readout elements of varying size, finely segmented both laterally and longitudinally. Such fine segmentation $\Delta\eta \times \Delta\phi = 0.025 \times 0.025$ in the second layer of the EM barrel, for example permits a detailed mapping of the electromagnetic and hadronic showers.

The position resolution of the EM is driven by the readout geometry (rectangular cells). There are three layers of cells, segmented along the particle’s direction of motion. The ϕ segmentation comes from grouping the accordion-shaped electrodes together into a common read out channel.

In the region $0 < |\eta| < 1.8$ the electromagnetic calorimeters are complemented by a “presampler” detector, an instrumented argon layer, which provides a measurement of the energy lost in the solenoid and the outer wall of the barrel cryostat.

The EMEC uses the same accordion geometry as the EMB, whereas the granularity is typically slightly larger than in the barrel.

The signal readout chain for the LAr calorimeter (indeed for all calorimeter systems) is divided into a fast analog readout for the trigger system and a slower digital readout used for more redefined trigger decisions and the offline reconstruction. However, regardless of the readout path, the signal

is initiated within the active LAr medium. To minimize noise and increase speed the first level of readout is located on the detector (both for LAr and Tile calorimeter, see 3.2.2). The front-end electronics amplify and shape the signal. Shaping electronics induce a bipolar pulse shape in the ionization signal. This shape is characterized by having both a positive and a negative component, which renders the integral of the signal exactly equal to zero.

The performance of the shaping electronics is critical for a correct energy calibration of the detector since the energy is primarily determined from the peak height of the pulse. In each calorimeter region, the overall pulse shape and duration are optimized to approximately cancel a constant injection of energy into the detector. The motivation for this approach is to effectively redefine the baseline of the energy measurement. In the high luminosity environment of the LHC, this reduces the sensitivity to the background from multiple pp interactions on average.

To translate these analog signals to digital signals that can be transmitted long distances to the next stage of the readout system, the pulse shape is measured over several 25 ns (nominal) time intervals, known as samples. The challenge of calorimeter calibration is to map these measured signals to the energy deposited in the active detector medium, known as the visible energy. This calibration is established using test-beam measurements of electrons in the EMB and EMEC calorimeters.

The hadronic calorimeter

Outside the EM calorimeter lies the system of hadronic calorimeters. The barrel portion, known as the Tile calorimeter, uses iron absorber slabs interspersed with scintillating tiles. The Tile calorimeter is most notable for its

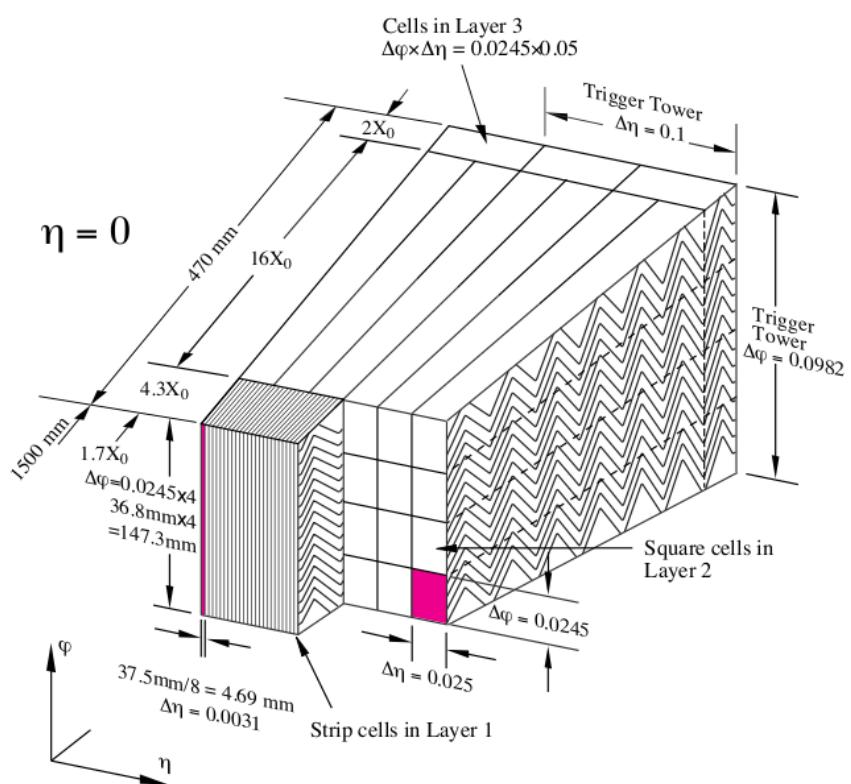


Figure 3.7: Cross section of the LAr barrel calorimeter where the different layers are visible. The granularity in η and ϕ of the cells of each of the three layers is also shown.

depth of 7.4 radiation lengths (λ^2). The hadronic end-cap and the forward calorimeter, which need to absorb the more energetic particles that are produced at large $|\eta|$, are made of copper and tungsten absorbers, respectively, with liquid argon as the active material.

The tile calorimeter is composed of 3 mm thick scintillating tiles, arranged to lie parallel to the incoming particle direction, interleaved with 14 mm thick iron plates. It is divided into the barrel calorimeter, covering $|\eta| < 1.0$, and two extended barrel calorimeters, covering $0.8 < |\eta| < 1.7$. Each tile is read out by two wavelength-shifting fibers, which convert the scintillator signal to visible light. The readout fibers of several tiles are grouped to a single photomultiplier tube forming cells in $\eta \times \phi$ space. As in the EM calorimeter, these cells are segmented into three layers, the first two of size $\Delta\eta = 0.1$ and $\Delta\phi = 0.1$ and the last of size $\Delta\eta = 0.2$ and $\Delta\phi = 0.1$. Towers to provide information to the trigger system are formed from 0.1×0.1 grouping of all three layers.

The HEC uses the LAr active readout design due to the higher radiation tolerance required for the forward regions. Although housed in the same cryostat as the accordion geometry EMEC, the HEC implements a flat-plate design.

The forward calorimeter extends to cover the region $3.1 < |\eta| < 4.9$. Since it is the only calorimeter that covers this very forward region, it must provide both electromagnetic and hadronic measurements. In addition, the high particle fluxes in this region necessitate a finely granulated design. The FCal is approximately 10 interaction lengths deep, and consists of three modules in

²To quantify the amount of material needed to capture a particle's energy, the unit of an interaction length, which is the distance over which a high energy charged particle loses $1 - \frac{1}{e} \sim 63\%$ of its energy, is commonly used.

each end-cap: the first, made of copper, is optimised for electromagnetic measurements, while the other two, made of tungsten, measure predominantly the energy of hadronic interactions.

The hadronic calorimeters are calibrated using muons in test-beam experiments and those muons produced by cosmic-rays in situ (see Chapter 4). The invariant mass of the Z boson in $Z \rightarrow ee$ events measured in-situ in the 2010 pp collisions is used to adjust the calibration derived from test-beams and cosmic-muons.

3.2.3 The Muon System

The muon system gives the ATLAS detector its overall shape and imposing nature, as depicted in Fig. 3.8. Muons have much smaller cross section to interact in material than electrons and hadrons, for this, they do not deposit all their energy in the calorimeters. The muon spectrometer is designed to detect muons within $|\eta| < 2.7$. Because many new physics signatures involve high-momentum muons, the system is also required to provide trigger signals based on the particle p_T for $|\eta| < 2.4$.

To provide a momentum measurement, the muons trajectories are bent in a toroidal magnetic field. This field is provided by one large barrel toroid and two large end-cap toroids, each toroid consisting of eight coils arranged symmetrically around the beam axis. The toroid system produces a magnetic field that is typically oriented in the ϕ direction and that is measured with over 1800 Hall sensors placed through the magnets. Under the influence of this field, muons are deflected in the $r-z$ plane and the transverse momentum of the muons is given then by the radius of curvature of the tracks. Since the highly-energetic muons bend very little even in this high magnetic field, the muon system is the largest of all the ATLAS sub-detectors, covering a

radius from ~ 4.5 m to ~ 12.5 m.

Four primary subsystems comprise the integrated muon spectrometer: monitored drift tubes (MDT), cathode strip chambers (CSC, which are multiwired proportional chambers with cathodes segmented into strips), resistive plate chambers (RPC) and thin gap chambers (TGC). The MDT and CSC subsystems are primarily designed for precision measurements of muon tracks, width the MDT system providing coverage for the more central region ($|\eta| < 2.7$, with full coverage only in $|\eta| < 2.0$), whereas the CSC is located in the more forward region ($2.0 < |\eta| < 2.7$) due to its ability to cope with higher background rates. The RPC and TGC muon subsystems are desinged to provide fast, robust readout for use in the trigger and data acquisition system. A detailed description of the subsystems can be found elsewhere [58].

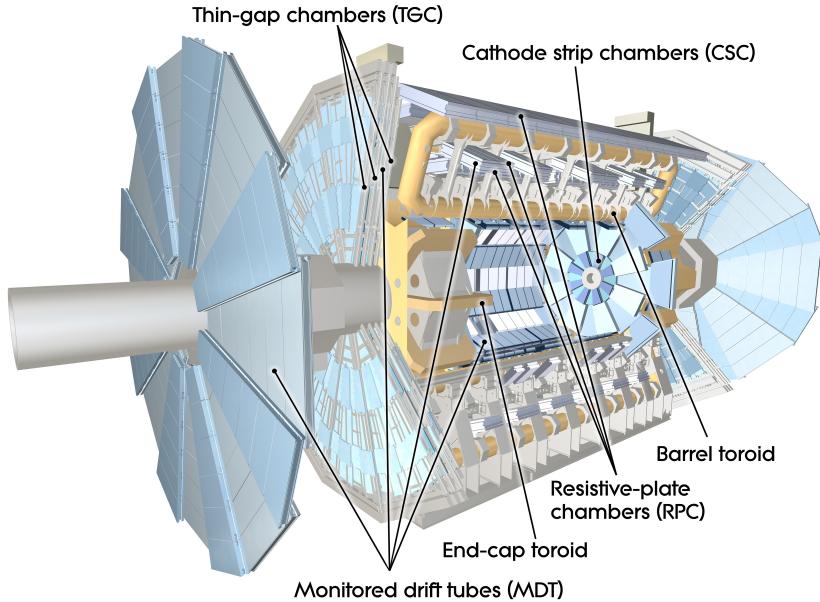


Figure 3.8: Muon Chamber

3.2.4 Forward detectors

Three smaller detector systems cover the ATLAS forward region. The main function of the first two systems is to determine the luminosity delivered to ATLAS. At ± 17 m from the interaction point lies LUCID (LUminosity measurement using Cerenkov Integrating Detector). The principle of LUCID is to detect inelastic $p - p$ scattering in the forward region, exploiting the fact that the number of particles detected is proportional to the total, both primary and pile-up, interactions in a bunch-crossing. LUCID thus provides a relative luminosity measurement, in which the detected number of particles must be translated to the total number of proton-proton interactions via calibration runs. The second detector is ALFA (Absolute Luminosity For ATLAS). Located at ± 240 m, it consists of scintillating fibre trackers located inside Roman pots which are designed to approach as close as 1 mm to the beam. The third system is Zero-Degree Calorimeter (ZDC), which place a role in determining the centrality of heavy-collisions.

3.2.5 Trigger and Data Adquisition

At design luminoisty, the LHC will deliver approximately 40 million collision events every second. With an average ATLAS event size of ~ 1.5 MB, this is far more information than can be saved into the finite data storage resources available. The goal of the trigger system is to move interesting physics events to permanent storage, while rejecting the vast majority of other events.

The online selection is done in three stages: the Level 1 (L1), Level 2 (L2) and Event Filter (EF) stages. Each trigger level refines the decisions made at the previous level and, where necessary, applies additional selection criteria. The data acquisition (DAQ) system receives and buffers the event data from

the detector-specific readout electronics, at the L1 trigger accept rate, over 1600 point-to-point readout links. The L1 trigger uses a limited amount of the total detector information (only from the calorimeter and the muon systems) using only simple hardware based algorithms to make a decision in less than $2.5 \mu\text{s}$, reducing the rate to about 75 kHz. The L2 and EF, collectively referred to as the High Level Trigger (HLT), are based on fast software algorithms running on large farms of commercial processors. The L2 is the first stage of the ATLAS DAQ system that has access to data from the ID and is capable of doing partial reconstruction of events up to the L1 accept rate. L2 trigger is designed to reduce the rate to approximately 3.5 kHz, with an event processing time of about 40 ms, averaged over all events. The EF reduces the rate to roughly 200 Hz. Its selections are implemented using offline analysis procedures within an average event processing time of the order of 4 s.

The L1 trigger is designed to accept high- p_T muons, electrons, photons, jets, and taus, as well as events with large missing transverse energy or sum energy. It uses signals from the TGCs and RPCs from muon triggers and reduced granularity calorimeter information for electron, photon, jet, tau and total energy triggers. The calorimeter trigger system, which maintains a fast readout independent from the remainder of the calorimeter is known as the Level-1 Calorimeter. At this level coarse calorimeter information is available in the form of jet elements with $\Delta\eta \times \Delta\phi = 0.2 \times 0.2$ for $|\eta| < 3.2$. Jets are reconstructed using a square sliding window algorithm. In addition to coarse jets, the total transverse energy is also measured at the L1. The region of the detector corresponding to the location where the L1 thresholds were passed – so called “region of interest” (RoI) – are then delivered to the L2 software algorithms.

The L2 trigger applies additional energy thresholds and multiplicity requirements using the RoI around triggered L1 objects. For example, the L2 jet trigger retrieves the data from cells surrounding the L1 RoI and constructs jets using a simplified cone jet algorithm.

The next step and last stage in the trigger chain is the EF, which receives events that have been selected by the L2 triggers and processes the entire event with the full detector granularity instead of only a restricted region.

The monitoring infrastructure of the HLT supports the real-time accumulation of histograms, and their aggregation across the farm, so that parameters can be extracted from cumulative distributions that contain events from all processor nodes. Beam parameters determined from those live histograms are transmitted online to the LHC and are also available to feed back into the HLT itself for use by its own trigger algorithms that depend on the precise knowledge of the luminous region.

3.2.6 ATLAS Performance and Data quality

The ATLAS detector has been operational for a number of years collecting large amounts of data. Before the start-up of the LHC, measuring muons from cosmic rays; which were used to test, understand, and align the detector. In 2010 and 2011 ATLAS recorded over 5.2fb^{-1} of collision data. Fig. 3.3 presented the luminosity delivered by the LHC in 2011 as well as the recorded luminosity by the detector, showing a good performance of the ATLAS Experiment. The fraction of time that each subdetector system was operational during data-taking is shown in Table 3.2.

The Data Quality (DQ) selection within ATLAS is based on the inspection of a standard set of distributions that lead to a data quality assessment which is encoded in so-called DQ flags. DQ flags are issued for each de-

tector, usually segmented in subdetectors like barrel, end-caps and forward. DQ flags are also issued for trigger slices and for each physics object reconstruction. In this way, the state of the ATLAS detector from hardware to physics object reconstruction is expressed through DQ flags, which are saved per luminosity block. A luminosity block is a time interval of typically two minutes.

The DQ information is used in analyses through dedicated lists of good runs/luminosity blocks. Good run lists are formed by DQ selection criteria in addition to other criteria, such as run range, magnetic field configuration and beam energy. A complete list of valid physics runs and luminosity blocks is used in each analysis.

Detector component	operational
Inner Detector	
Pixel	≈96.4%
SCT	≈99.2%
TRT	≈97.5%
Calorimeter	
EM	≈99.8%
Tile	≈96.2%
Hadronic, end-cap	≈99.6%
Forward calorimeter	≈99.8%
Muon Spectrometer	
MDT	≈99.7%
CSC	≈97.7%
RPC	≈97.0%
TGC	≈97.9%

Table 3.2: The approximate fraction of time that each individual subdetector system was operational during data-taking.

Chapter 4

Event reconstruction and *b*-Tagging

The event reconstruction software, which in ATLAS is implemented in the software framework ATHENA [62], process the events starting from the raw data obtained from the various sub-detectors (energy deposits and hits), processing them in many different stages and finally interpreting them as a set of charged tracks, electrons, photons, jets, muons and, in general, of possible kinds of final state objects with related four momenta. In this chapter the reconstruction of these objects is briefly described together with the algorithms for the identification of *b*-quark jets. These algorithms are mainly based on the reconstruction of the primary interaction vertex, on the reconstruction of charged particles in the Inner Detector and on the reconstruction of jets in the calorimeter.

4.1 Jet reconstruction and calibration

Hadronic jets used for ATLAS analyses are reconstructed by a jet algorithm, starting from the energy depositions of electromagnetic and hadronic showers in the calorimeters. Two different size parameters are used: $R = 0.4$, for narrow jets, and $R = 0.6$, for wider jets. The default jet algorithm is the anti- k_t algorithm, described in Chapter 2. Due to the expected level of pile-up in the LHC, the primary factor that influenced the selection of this algorithm was the effect of multiple simultaneous interactions on the reconstruction of jets. The original ATLAS cone algorithm, known to contain infrared and collinear sensitivity, is highly susceptible to this effect. On the contrary, the anti- k_t algorithm is the most stable after the introduction of pile-up [63].

The input to calorimeter jet reconstruction can be calorimeter towers or topological cell clusters. Charged particle tracks reconstructed in the Inner Detectors are also used to define jets. The latter have the further advantage of being insensitive to pile-up and they provide a stable reference for systematic studies. The jet inputs are combined as massless four-momentum objects in order to form the final four-momentum of the jet, which allows for a well-defined jet mass [64]. In the case of track-jets, the track four-momentum is constructed assuming the π meson mass for each track.

Calorimeter towers are static, $\Delta\eta \times \Delta\phi = 0.1 \times 0.1$, grid elements built directly from calorimeter cells. There are two types of calorimeter towers: with or without noise suppression. The latter are called “noise-suppressed” towers and use only the cells with energies above a certain noise threshold. The noise of a calorimeter cell is measured by recording calorimeter signals in periods where no beam is present in the acelerator. The standard deviation σ around the mean measured energy is interpreted as the noise of the cell, and dependes on the sampling layer in which the cell resides and the position

in η .

The results presented in this thesis show jets which were built from noise-suppressed topological clusters of energy in the calorimeter, also known as “topo-clusters” [65]. Topological clusters are groups of calorimeter cells that are designed to follow the shower development taking advantage of the fine segmentation of the ATLAS calorimeters. The topological cluster formation starts from a seed cell with $|E_{cell}| > 4\sigma$ above the noise. In a second step, neighbor cells that have an energy at least 2σ above their mean noise are added to the cluster. Finally, all nearest-neighbor cells surrounding the clustered cells are added to the cluster, regardless of signal-to-noise ratio¹. The position of the cluster is assigned as the energy-weighted centroid of all constituent cells (the weight used is the absolute cell energy).

In Monte Carlo simulation, reference jets (“truth jets”) are formed from simulated stable particles using the jet algorithm utilized for the reconstructed jets.

Jet calibration

The purpose of the jet energy calibration, or jet energy scale (JES), is to correct the measured electromagnetic scale (EM scale) energy to the energy of the stable particles within a jet. The jet energy calibration must account then for the calorimeter non-compensation; the energy lost in inactive regions of the detector, such as in the cryostat walls or cabling; energy that escapes the calorimeters, such as that of highly-energetic particles that “punch-through” to the muon system; energy of cells that are not included in clusters, due to inefficiencies in the noise-suppression scheme; and energy of clusters not

¹Noise-suppressed towers also make use of the topological clusters algorithm [65] to select cells, i.e. only calorimeter cells that are included in topo-clusters are used.

included in the final reconstructed jet, due to inefficiencies in the jet reconstruction algorithm. The muons and neutrinos that may be present within the jet are not expected to interact within the calorimeters, and are not included in this energy calibration. Due to the varying calorimeter coverage, detector technology, and amount of upstream inactive material, the calibration that must be applied to each jet to bring it to the hadronic scale varies with its η position within the detector.

The jet energy is first reconstructed from the constituent cell energies at EM scale. These cells have been calibrated to return the energy corresponding to electromagnetic showers in the calorimeter, based on test-beam injection of electrons and pions [66], measurements of cosmic muons [67] and the reconstruction of the Z mass peak in $Z \rightarrow ee$ decays [68]. The correction for the lower response to hadrons is based on the topology of the energy depositions observed in the calorimeter.

In the simplest case the measured jet energy is corrected, on average, using Monte Carlo simulations, as follows:

$$E_{calib}^{jet} = E_{meas}^{jet}/F_{calib}(E_{meas}^{jet}), \quad \text{with } E_{meas}^{jet} = E_{EM}^{jet} - O(\text{NPV}), \quad (4.1)$$

where E_{EM}^{jet} is the calorimeter energy measured at the electromagnetic scale, E_{calib}^{jet} is the calibrated energy and F_{calib} is the calibration function that depends on the measured jet energy and is evaluated in small jet η regions. The variable $O(\text{NPV})$ denotes the correction for additional energy from multiple proton-proton interactions depending on the number of primary vertices (NPV).

The simplest calibration scheme and the one used in this thesis is the so called “EM+JES”. This calibration applies the corrections as a function of the jet energy and pseudorapidity to jets reconstructed at the electromagnetic scale. The additional energy due to multiple proton-proton collisions within

the same bunch crossing (pile-up) is corrected before the hadronic energy scale is restored, such that the derivation of the jet energy scale calibration is factorised and does not depend on the number of additional interactions in the event. The EM+JES calibration scheme consists of three subsequent steps:

- Pile-up correction: An offset correction is applied in order to subtract the additional average energy measured in the calorimeter due to multiple proton-proton interactions. This correction is derived from minimum bias data as a function of NPV, the jet pseudorapidity and the bunch spacing.
- Vertex correction: The jet four momentum is corrected such that the jet originates from the primary vertex of the interaction instead of the geometrical centre of the detector.
- Jet energy and direction correction: The jet energy and direction are corrected using constants derived from the comparison of the kinematic observables of reconstructed jets and those from truth jets in the simulation.

In the final step the calibration is derived in terms of the energy response of the jet, or the ratio of the reconstructed jet energy to that of a truth jet. The EM scale response is written as,

$$R_{EM}^{jet} = E_{EM}^{jet}/E_{truth}^{jet} \quad (4.2)$$

To compute this quantity, reconstructed jets must be matched to isolated jets in the Monte Carlo within $\Delta R < 0.3$. The isolation requirement is applied in order to factorize the effects due to close-by jets from those due

to purely detector effects such as dead material and non-compensation. The isolation criterion requires that no other jet with a $p_T > 7$ GeV be within $\Delta R < 2.5R$, where R is the distance parameter of the jet algorithm. The EM scale energy response is binned in truth jet energy, E_{truth}^{jet} , and the calorimeter jet detector η . For each (E_{truth}^{jet}, η) -bin, the averaged jet response is defined as the peak position of a Gaussian fit to the $E_{EM}^{jet}/E_{truth}^{jet}$ distribution. A function $F_{calib,k}(E_{EM}^{jet})$ is then defined for each η -bin k that describes the response as a function of the uncalibrated jet energy. $F_{calib,k}(E_{EM}^{jet})$ is parameterised as:

$$F_{calib,k}(E_{EM}^{jet}) = \sum_{i=0}^{N_{max}} a_i (\ln E_{EM}^{jet})^i, \quad (4.3)$$

where a_i are free parameters, and N_{max} is chosen between 1 and 6 depending on the goodness of the fit. The final jet energy scale correction that relates the measured calorimeter jet energy scale to the hadronic scale is then defined as $1/F_{calib,k}(E_{EM}^{jet})$ in the following:

$$E_{EM+JES}^{jet} = \frac{E_{EM}^{jet}}{F_{calib}(E_{EM}^{jet})|_\eta}, \quad (4.4)$$

where $F_{calib}(E_{EM}^{jet})|_\eta$ is $F_{calib,k}(E_{EM}^{jet})$ for the relevant η -bin k .

Other calibrations schemes are the global calorimeter cell weighting (GCW) calibration and the local cluster weighting (LCW) calibration. The GCW scheme exploits the observation that electromagnetic showers in the calorimeter leave more compact energy depositions than hadronic showers with the same energy. Energy corrections are derived for each cell within a jet. The cell corrections account for all energy losses of a jet in the detector. Since these corrections are only applicable to jets and not to energy depositions, they are called “global” corrections.

The LCW calibration method first classifies topo-clusters as either electromagnetic or hadronic, based on the measured energy density. Energy

corrections are derived according to this classification from single charged and neutral pion Monte Carlo simulations. Dedicated corrections are derived for the effects of non-compensation, signal losses due to noise threshold effects, and energy lost in non-instrumented regions. Since the energy corrections are applied without reference to a jet definition they are called “local” corrections. Jets are then built from these calibrated clusters using a jet algorithm.

The final jet energy calibration can be applied to EM scale jets, with the resulting calibrated jets referred to as EM+JES, or to LCW (GCW) calibrated jets, with the resulting jets referred to as LCW+JES (GCW+JES) jets.

A further jet calibration scheme called global sequential (GS) calibration, starts from jets calibrated with the EM+JES calibration and exploits the topology of the energy deposits in the calorimeter to characterise fluctuations in the jet particle content of the hadronic shower development. Correcting for such fluctuations can improve the jet energy resolution. The correction uses several jet properties, and each correction is applied sequentially.

For the 2011 data the recommended calibration schemes were the EM+JES and the LCW calibrations. The simple EM+JES calibration does not provide the best performance, but allows in the central detector region the most direct evaluation of the systematic uncertainties from the calorimeter response to single isolated hadrons measured *in situ* and in test-beams and from systematic variations in the Monte Carlo simulation. For the LCW+JES calibration scheme the JES uncertainty is determined from *in situ* techniques. For all calibration schemes, the JES uncertainty in the forward regions is derived from the uncertainty in the central region using the transverse momentum balance in events where only two jets are produced.

Jet energy scale uncertainties for the EM+JES scheme

For many physics analyses, the uncertainty on the JES constitutes the dominant systematic uncertainty because of its tendency to shift jets in and out of analysis selections due to the steeply falling jet p_T spectrum. The uncertainty on the EM+JES scale is determined primarily by six factors: varying the physics models for hadronization and parameters of the Monte Carlo generators, evaluating the baseline calorimeter response to single particles, comparing multiple models for the detector simulation of hadronic showers, assessing the calibration scales as a function of pseudorapidity, and by adjusting the JES calibration methods itself. The final JES uncertainty in the central region, $|\eta| < 0.8$, is determined from the maximum deviation in response observed with respect to the response in the nominal sample. For the more forward region, the so called “ η -intercalibration” contribution is estimated. This is a procedure that uses direct di-jet balance measurements in two-jet events to measure the relative energy scale of jets in the more forward regions compared to jets in a reference region. The technique exploits the fact that these jets are expected to have equal p_T due to transverse momentum conservation. Figure 4.1 shows the final fractional jet energy scale uncertainty and its individual contributions as a function of p_T for a central η region. The JES uncertainty for anti- k_t jets with $R = 0.4$ is between $\approx 4\%$ (8%, 14%) at low jet p_T and $\approx 2.5\%-3\%$ (2.5%-3.5%, 5%) for jets with $p_T > 60$ GeV in the central (endcap, forward) region.

In addition to the tests above, *in situ* tests of the JES using direct γ -jet balance, multi-jet balance, and track-jets indicate that the uncertainties in Fig. 4.1 reflect accurately the true uncertainties in the JES.

In the case of jets induced by bottom quarks (b -jets), the calorimeter response uncertainties are also evaluated using single hadron response mea-

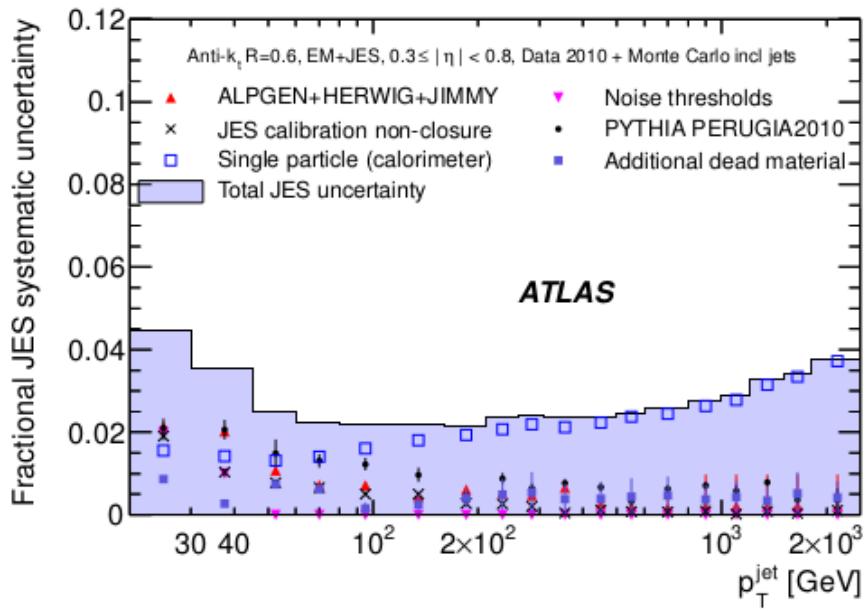


Figure 4.1: Fractional jet energy scale uncertainty as a function of jet p_T for jets in the pseudorapidity region $0.3 < |\eta| < 0.8$ in the calorimeter barrel. The total uncertainty is shown as the solid tight blue area. The individual sources are also shown.

surements *in situ* and in test beams [69]. For jets within $|\eta| < 0.8$ and $20 \leq p_T < 250$ GeV the expected difference in the calorimeter response uncertainty of identified b -jets with respect to the one of inclusive jets is less than 0.5%. It is assumed that this uncertainty extends up to $|\eta| < 2.5$.

The JES uncertainty arising from the modelling of the b -quark fragmentation can be determined from systematics variations of the Monte Carlo simulation. The fragmentation function is used to estimate the momentum carried by the b -hadron with respect to that of the b -quark after quark fragmentation. The fragmentation function included in PYTHIA originates from a detailed study of the b -quark fragmentation function in comparison with OPAL [70] and SLD [71] data. To assess the impact of the b -fragmentation, the nominal parameters of the PYTHIA fragmentation function are replaced by the values from a tune using the Professor framework [72]. In addition, the nominal fragmentation function is replaced by the modified Bowler-Lund fragmentation function [73]. The b -jet response uncertainty is evaluated from the ratio between the response of b -jets in the varied Monte Carlo samples to the nominal PYTHIA. The response variations are well within 2%.

The b -jet JES uncertainty is obtained adding the calorimeter response uncertainty and the uncertainties from the systematic Monte Carlo variations in quadrature. The resulting additional JES uncertainty for b -jets is shown in Fig. 4.2. It is about 2% up to $p_T \approx 100$ GeV and below 1% for higher p_T . To obtain the overall b -jet uncertainty this uncertainty is added in quadrature to the JES uncertainty for inclusive jets.

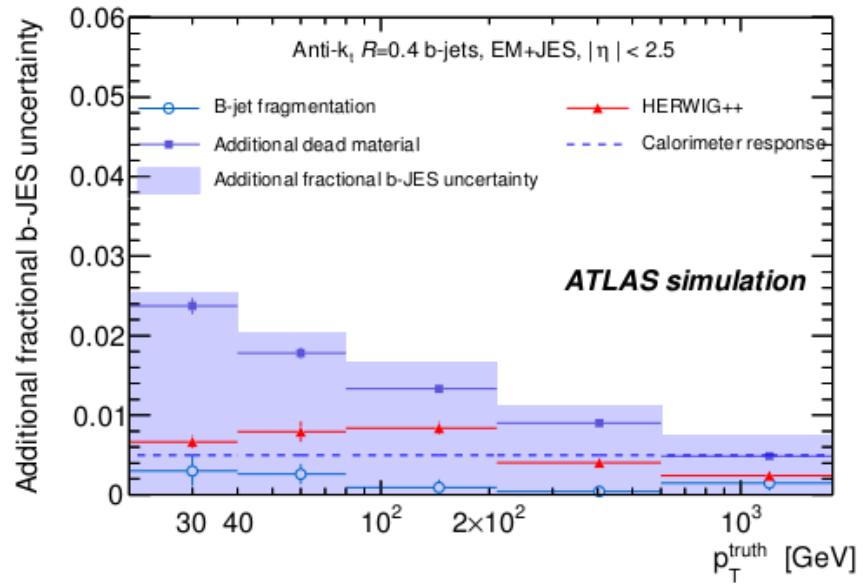


Figure 4.2: Additional fractional b -jet JES uncertainty as a function of the truth jet transverse momentum for anti- k_t jets with $R = 0.4$ calibrated with EM+JES scheme for $|\eta| < 2.5$. Shown are systematic Monte Carlo variations using different modelling of the b -quark fragmentation and physics effects as well as variations in the detector geometry and the uncertainty in the calorimeter response to b -jets as evaluated from single hadron response measurements. Uncertainties in the individual points are statistical only.

4.2 Reconstruction of charged particle tracks

The Inner Detector layout and the characteristics of its main sub-detectors were presented in Section 3.2.1 of Chapter 3. The tracking algorithm is based on a modular software framework, which is described in more detail in Ref. [74]. The main steps of the tracking algorithm are the following:

- Firstly, the raw data from the pixel and SCT detectors are converted into clusters, while the TRT raw timing information is turned into calibrated drift circles. The SCT clusters need to be further transformed into space-points, by combining the clusters information from opposite sides of the SCT module (stereo strip layers).
- In a second stage, the track-finding is performed, in which the pattern recognition and a global χ^2 minimization procedure is implemented as a default.

In the track-finding stage, track seeds are found in the first three pixel layers and in the first SCT layer. These are extended throughout the SCT to form track candidates and a first track fit is performed. Afterwards, ambiguities in the track candidates found in the silicon detectors are resolved, and tracks are extended into the TRT (which covers up to $|\eta| < 2$). The final track candidate is refitted with the full information from the three tracking subdetectors. The baseline algorithm is designed for the efficient reconstruction of primary charged particles. Primary particles are defined as particles with a meanlife of greater than 3×10^{-11} s directly produced in a proton-proton interaction, or from the subsequent decays or interactions of particles with lifetime shorter than 3×10^{-11} s. The tracks reconstructed in this stage are required to have $p_T > 400$ MeV.

In a complementary stage, a track search starts from segments reconstructed in the TRT and extends them inwards by adding silicon hits, which is referred to as “back-tracking”. This recovers tracks for which the first hits in the pixel layers are missing, e.g. because they originate from secondaries, which are produced in the interaction of primaries.

The final reconstructed track trajectory is usually defined at its closest point to the interaction region on the transverse plane by its impact parameters in the transverse plane and in the longitudinal direction, respectively called d_0 and z_0 ², and by its momentum, typically expressed in azimuthal angle ϕ , polar angle θ and inverse momentum $1/p$.

The track reconstruction efficiency is defined as the fraction of primary particles with $p_T > 400$ MeV and $|\eta| < 2.5$ matched to a reconstructed track. The reconstruction efficiency for primary tracks with transverse momentum above 1 GeV and central η is above 80%, going down to values below 70% for tracks at the edge of the Inner Detector acceptance [59]. The dense environment of a jet decreases the track reconstruction efficiency and increases the fake rate. This is caused by the occurrence of shared hits between different tracks, which makes the pattern recognition and track fitting tasks more difficult.

The relative transverse momentum scale and resolution of tracks is defined as the Gaussian mean and width of

$$p_T^{MC} \times (1/p_T^{MC} - 1/p_T^{reco}) = 1 - \frac{p_T^{MC}}{p_T^{reco}} \quad (4.5)$$

where $p_T^{MC,reco}$, refers to the track’s transverse momentum given by simulation truth (MC) or by reconstruction (reco). It should be noted that the $(1/p_T)$ resolution is used instead of $\sigma(p_T)$ as the Inner Detector measures the

²Strickly speaking the impact parameter is $|z_0| \sin \theta$, where θ is the polar angle of the track.

sagitta and not directly the transverse momentum³. However, the resolution obtained from the equation above is the relative transverse momentum resolution, $\sigma(p_T)/p_T$. At low p_T the multiple scattering dominates the resolution, and at high momenta, the resolution is limited by the bending power of the solenoid field and by the intrinsic detector resolution. For a central track with $p_T = 5$ GeV the transverse momentum resolution is around 75 MeV and the transverse impact parameter resolution is about $35 \mu\text{m}$.

4.3 Vertex reconstruction

Primary vertices are reconstructed using an iterative vertex finding algorithm [75]. In a first step, a dedicated vertex finding algorithm associates tracks to vertex candidates. Vertex seeds are obtained by looking for the global maximum in the distribution of the z coordinates of the tracks. In a second stage, an iterative χ^2 fit is made using the seed and nearby tracks. Each track carries a weight which is a measure of its compatibility with the fitted vertex depending on the χ^2 of the fit. Tracks displaced by more than 7σ from the vertex are used to seed a new vertex and the procedure is repeated until no additional vertices can be found. The parameters of the beam spot are used both during the finding to preselect compatible tracks and during the fitting step to constrain the vertex fit.

The knowledge of the position of the primary interaction point (primary vertex) of the proton-proton collision is important for b -quark jets identification since it defines the reference point with respect to which impact parameters and vertex displacements are measured. The typical vertexing resolution in z is $O(100\mu\text{m})$.

³The relation between sagitta s and transverse momentum (p_T) is given by $s \sim 1/p_T$.

To ensure a good resolution on the vertex position, the primary vertex must be reconstructed from at least five tracks. The choice of the primary vertex is less trivial in the presence of minimum-bias events from pile-up: the primary vertex from a pile-up event may be mistakenly used as the signal vertex, or a fake primary vertex built from tracks from two different vertices may be reconstructed. The current strategy is to choose the primary vertex candidate that maximizes $\sum_{tracks} p_T^2$.

4.4 b -jet Tagging

The ability to identify jets originating from *bottom*-quarks (denoted as b -tagging in the following) is important for the high- p_T physics program of a general-purpose experiment at the LHC such as ATLAS since many interesting physics processes contain b -quarks in the final state, while the most abundant backgrounds contain mostly up, down and strange quark or gluon jets or, in a smaller fraction of cases, charm quark jets. The aim of b -tagging is therefore to identify the b -quark jets with high efficiency, while rejecting most of the background contamination from jets originating from the fragmentation of light (u , d , and s) quarks, gluons and c -quarks.

A b -quark, once produced, fragments necessarily into a b -flavoured hadron, b -hadron in the following. In most of the cases ($\approx 87\%$), first an excited b -hadron, like a B^* or a B^{**} , which decays immediately, strongly or electromagnetically, into a ground state b -hadron plus one or more further particles; while in the remaining cases, a ground state b -hadron is produced directly. One is only interested in the transition from a b -quark into the final state b -hadron, since the typical timescale for electromagnetic or strong interactions is so small that the B^* , B^{**} decay vertices are not significantly displaced

with respect to the primary vertex. In most of the cases ($\approx 91\%$) a b -meson is produced out of the fragmentation of an original b -quark (40% B^+ , 40% B^0 and 11% B_s^0). The rest are b -baryons.

Due to the b -quark fragmentation function being very hard, most of the original b -quark energy is transmitted to the final b -hadron. This fraction is for example 70% for b -quarks with a momentum of ≈ 45 GeV. This property can be exploited during b -tagging, since the fragmentation for light quarks into light hadrons or c -quarks into c -hadrons is softer.

Any of the finally produced b -hadrons decay through weak interactions and therefore have a significant lifetime, which is on average, for all b -hadrons considered, $(1.568 \pm 0.009) \times 10^{-12}$ s. The effective distance travelled in the detector by the b -hadron before decaying depends on the b -hadron momentum, which enters the relativistic boost factor $\beta\gamma$. A b -quark with momentum of 50 GeV will travel around 3 mm, which is a visible flight length in the detector. Due to the combination of the b -hadron lifetime and relatively high mass ($m_B \approx 5.28$ GeV), which results in a non-negligible decay angle of the b -hadron decay products with respect to the b -hadron flight direction, the charged particles produced at the decay vertex will be on average significantly displaced with respect to the primary vertex position.

This is the main signature which is exploited by the *lifetime* based b -tagging algorithms, which are based either on the presence of significantly displaced tracks, as in impact parameter based b -tagging algorithms, or on the explicit reconstruction of the b -hadron decay vertex, as in secondary vertex based b -tagging algorithms.

b -hadrons decay preferably into a c -hadron plus additional particles⁴. The lifetime of a c -hadron is not much lower than for b -hadrons, but in general

⁴Weak decays are governed by the CKM matrix mechanism, and $|V_{cb}|^2 \gg |V_{ub}|^2$.

the momentum of the c -hadron will be lower than the original b -hadron momentum. However, the c -hadron can still travel for a significant path in the detector and form with its decay products a visible *tertiary* vertex.

Another property which is usually exploited by b -tagging is the fraction of b - and c -hadron decays into leptons: a lepton from the semi-leptonic decay of a b -hadron ($b \rightarrow l$) or from the subsequent c -hadron decay ($b \rightarrow c \rightarrow l$) is produced in a b -quark in $\approx 21\%$ of the cases. This is valid both in case the lepton is an electron or a muon, which brings the overall fraction of b -quarks ending up into a lepton to $\approx 42\%$. Due to the b - or c -hadron mass, the lepton will be emitted with an average transverse momentum comparable with m_{b-had} or m_{c-had} . By identifying either an electron or a muon originating from a jet and by requiring it to have sufficiently high p_T with respect to the jet axis, it is possible to identify b -jets.

Association of tracks to jets

The b -tagging performance relies critically on the accurate reconstruction of the charged tracks in the ATLAS Inner Detector. The actual tagging is performed on the sub-set of tracks in the event that are associated to jets. The b -tagging algorithm takes as input the three-momenta of the jets, reconstructed by a jet algorithm, and uses the jet direction to associate the charged particles tracks to the jet. Since the 2 Tesla solenoidal magnetic field of the ATLAS Inner Detector bends charged particles in the transverse plane, in particular in the case of low p_T tracks, the tracks are best matched to the jet by using the direction of their momenta at the point of closest approach to the interaction region. The criterion for associating charged particle tracks to jets is simply:

$$\Delta R(jet, track) < \Delta R_{cut} \quad (4.6)$$

where usually the value of $\Delta R_{cut} = R$ is used; with R , the distance parameter of the jet algorithm used for jet reconstruction.

After the tracks are associated to the jets, they are filtered in order to remove tracks with bad quality or which can easily be erroneously identified as secondary tracks from b -decays. These include tracks originating from decays of even longer lived particles, like K_s^0 ($c\tau \approx 2.69$ cm) and Λ baryons ($c\tau \approx 7.89$ cm); from electromagnetic interactions in the detector material, like conversions in electron-positron pairs ($\gamma \rightarrow e^+e^-$); or from hadronic interactions with the detector material, which result in two or more tracks with high impact parameter. In order to reject badly reconstructed tracks, quality cuts are applied. Requirements are imposed on the number of silicon hits, the track fit quality, the track momentum, and the transverse and longitudinal impact parameters. The track selection needs to be particularly tight in the case of the impact parameter based b -tagging algorithms, since in that case the explicit presence of a vertex is not required, so that the influence of badly reconstructed tracks or tracks from long lived particles does directly limit the performance. The minimum track p_T required is of 1 GeV in the case of the impact parameter based algorithms and of 400-500 MeV otherwise. The transverse and longitudinal impact parameters must fulfill $|d_0| < 1$ mm (3.5 mm) and $|z_0| \sin \theta < 1.5$ mm (no cut on z_0) in the case of the algorithms relying on the impact parameters of tracks (on the reconstruction of secondary vertices). The minimum number of precision hits required is typically of 7 hits, for both approaches.

4.4.1 b -tagging algorithms

For the 2011 data-taking a set of lifetime taggers were commissioned and calibrated. In this section a brief description of the main features of these

algorithms will be given.

Impact parameter based b -tagging algorithms

The charged particle tracks originating from b -hadrons are expected to have significantly higher transverse and longitudinal impact parameters compared to prompt tracks originating directly from fragmentation. If the effect of long lived particles, conversions and hadronic interactions can be reduced, the best discrimination between prompt tracks and displaced tracks from b - and c -hadron decays can be obtained using the impact parameter significance, both in the transverse and longitudinal plane. Being,

$$IP_{r\phi} = d_0 \text{ and } IP_z = z_0 \sin \theta, \quad (4.7)$$

The transverse and longitudinal impact parameter significances are obtained by dividing $IP_{r\phi}$ and IP_z by their respective errors,

$$IP_{r\phi}/\sigma(IP_{r\phi}) \text{ and } IP_z/\sigma(IP_z). \quad (4.8)$$

On the basis that the decay point of the b -hadron must lie along its flight path, and in order to increase the discriminating power of the impact parameter significance, a lifetime sign is assigned to these variables (replacing the sign of the geometrical definition of the impact parameter). The sign is positive if the track extrapolation crosses the jet direction in front of the primary vertex (i.e. is more compatible with having its origin in a secondary decay vertex in the direction of flight expected for the b -hadron) or negative if the track is more likely to intersect the flight axis behind the primary vertex, opposite to the jet direction. Both cases are illustrated in Fig. 4.3.

The lifetime sign can be defined in three-dimensions, according to the variables $\vec{p}_{T_{jet}}$, $\vec{p}_{T_{trk}}$ and $\vec{\Delta r}_{IP} = \vec{r}_{IP} - \vec{r}_{PV}$, the three-dimensional impact

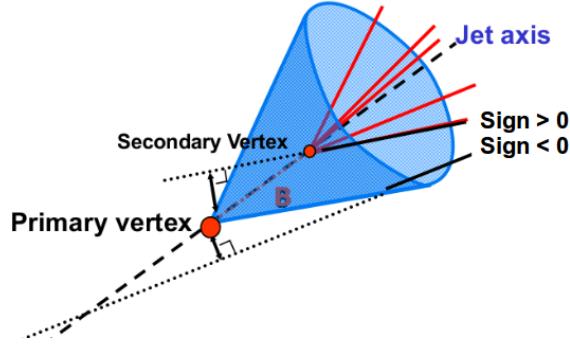


Figure 4.3: Lifetime sign of tracks. A positive and a negative lifetime signed track is shown.

parameter of the track with respect to the primary vertex:

$$\text{sign}_{3D} = \text{sign}([\vec{p}_{trk} \times \vec{p}_{jet}] \cdot [\vec{p}_{trk} \times \vec{\Delta r}_{IP}]). \quad (4.9)$$

The computation of the lifetime sign assumes that the jet direction reproduces, up to a good approximation, the b -hadron direction. Under this assumption and up to resolution effects both on the jet direction and on the impact parameter and momentum of the track, the lifetime sign of tracks originating from b -hadron decays is positive.

The lifetime sign can also be defined on the transverse plane ($x - y$) or on the longitudinal plane ($r\phi - z$) by considering respectively the transverse and longitudinal impact parameters (the projections of the three-dimensional impact parameter on the respective planes):

$$\text{sign}_{r\phi} = \text{sign}(\sin(\phi_{jet} - \phi_{trk}) \cdot d_{0,trk}); \text{ and } \text{sign}_z = \text{sign}((\eta_{jet} - \eta_{trk}) \cdot z_{0,trk}). \quad (4.10)$$

Tracks from the fragmentation in light-jets tend to have a signed impact parameter distribution which is symmetric around 0, since they have no

correlation with the jet direction. Tracks from b - and c -hadron decays, as expected, have an asymmetric distribution, with the most significant contribution at positive significances; however a negative tail extending beyond the pure fragmentation contribution is also seen, corresponding to resolution effects and to an eventual mismatch between the b -jet and the b -hadron directions.

The significance, which gives more weight to tracks measured precisely, is the main ingredient of the tagging algorithms based on impact parameters. Now, the impact parameter significance of all N tracks associated to the jet to tag need to be combined into a single discriminating variable. It is assumed that tracks are uncorrelated, so their probability density functions (PDF), defined based on the transverse and/or longitudinal impact parameter significance distributions for the different hypothesis, are uniquely defined as a function of the jet flavour. Using a likelihood function defined according to the product of these PDFs, under the hypothesis of uncorrelated tracks, the following likelihood ratio provides the optimal separation, according to Neyman-Person lemma [76]:

$$\text{LR}(IP_1, IP_2, \dots, IP_N) = \frac{\prod_{i=1}^N \text{PDF}_b(IP_i)}{\prod_{i=1}^N \text{PDF}_l(IP_i)} \quad (4.11)$$

For convention, the discriminant variable used for b -tagging is then defined as:

$$\text{weight}(IP_1, IP_2, \dots, IP_N) = \log(\text{LR}(IP_1, IP_2, \dots, IP_N)) \quad (4.12)$$

Using such a formalism, two impact parameter based b -tagging algorithms are constructed, based on the definition of $\text{PDF}(IP_i)$:

- 1. IP2D: $\text{PDF}(IP_i) = \text{PDF}(IP_{i,r\phi})$
- 2. IP3D: $\text{PDF}(IP_i) = \text{PDF}(IP_{i,r\phi}, IP_{i,z})$

In the first case the track PDF is one-dimensional, based on the transverse impact parameter significance. In the second case the PDF is based on a two-dimensional histogram of the transverse and longitudinal impact parameter significance.

The **IP3D** is one of the high-performance tagging algorithms supported for the 2011 data-taking, in which input variables are compared to pre-defined smooth Monte Carlo PDFs for both b -jet and light jet hypotheses [77]. Prior to the use of these advanced tagger, a simpler tagging algorithm, the **Jet-Prob**, combining the impact parameter significances of all tracks associated to the jet was devised to be used for early data, being extensively used during 2010 [78].

The impact parameter based algorithm permits to obtain a very good b -tagging performance, as will be shown at the end of this chapter. This performance can be improved by using some information from the secondary vertex based vertexing algorithms in two aspects: tracks associated to long lived particle vertices can be removed from the tracks considered for the impact parameter based algorithms; and, the direction between the secondary and the primary vertex positions can be used to improve the reliability of the lifetime sign, substituting \vec{p}_{jet} with $\vec{r}_{SV} - \vec{r}_{PV}$. The latter improves significantly the estimation of the b -hadron direction. Both kinds of information improve the performance of the impact parameter based b -tagging algorithms.

Secondary vertex based b -tagging algorithms

The typical topology of particle decays in a b -jet is a decay chain with two vertices, one stemming from the b -hadron decay and at least one from c -hadron decays. The reconstruction of these secondary vertices is done in an inclusive way, where the number of charged particle tracks originating from the b - and

c -hadron decays is not known a-priori. An exclusive reconstruction of the huge number of different possible b -decay modes cannot be performed, the set of selection cuts needed to reconstruct all of them would severely limit the reconstruction efficiency.

Two strategies to detect a secondary decay vertex in b -jets are available in ATLAS. The first one is based on the fit of a single geometrical vertex. Even if this hypothesis is not correct, this approximation works well for a large fraction of cases. The second algorithm is based on a kinematic approach, which assumes that the primary event vertex and the b - and the c -hadron decay vertices lie approximately on the same line, the flight path of the b -hadron.

The inclusive fit of a single displaced vertex in b -jets is based on the VKalVrt [79] reconstruction package. The main idea of the algorithm implemented in this package is to maximise the b/c -hadron vertex detection efficiency, keeping at the same time the probability to find a vertex inside a light jet low.

The algorithm starts with all tracks associated to the jet and passing a loose cut selection. The vertex search starts with looking for all track pairs and trying to form a two-track vertex. Each track of the pair must have a three-dimensional impact parameter significance with respect to the primary vertex larger than 2σ and the sum of these two significances must be larger than 6σ . To reduce the influence of badly measured tracks, the two-tracks vertices are required to be produced in the direction of flight of the b -quark, by requiring the scalar product of $(\vec{r}_{2-track} - \vec{r}_{PV}) \cdot \vec{p}_{jet}$ to be positive. Charged particles coming from long lived particles and conversions are not considered for the inclusive b -decay vertex fit. All the tracks corresponding to the accepted two-track vertices are used to determine a single secondary

vertex. If the resulting vertex has a very small vertex probability, the track with the highest contribution to the vertex χ^2 is removed from the vertex fit and the vertex fit is repeated until the χ^2 of the vertex fit is good. The result of this procedure is the (eventual) presence of a vertex, its position, and the list of its associated tracks.

The **SV1** secondary vertex algorithm uses this procedure to reconstruct inclusive secondary vertices. This advanced tagger takes advantage of three of the reconstructed vertex properties: the invariant mass of all tracks associated to the vertex, the ratio of the sum of the energies of the tracks in the vertex to the sum of the energies of all tracks in the jet, and the number of two-track vertices. These variables are combined using a likelihood ratio technique. SV1 relies on a two-dimensional distribution of the two first variables and a one-dimensional distribution of the number of two-track vertices. In addition the distance ΔR between the jet axis and the line joining the primary vertex to the secondary one is used.

The three-dimensional decay length significance alone, signed with respect to the jet direction can be used as a discriminating variable between b -jets and light jets: this is the principle of the early data **SV0** tagger, extensively used as well with the 2010 and 2011 data [80].

As opposed to the algorithm described above, in which the displaced tracks are selected and an inclusive single vertex is obtained, a second algorithm, called **JetFitter**, is based on a different hypothesis. It assumes that the b - and the c -hadron decay vertices lie on the same line defined through the b -hadron flight path. All charged particle tracks stemming from either decay intersect this b -hadron flight axis. This method has the advantage of reconstructing incomplete topologies, with, for instance, a single track from the b -hadron and a single track from the c -hadron decay. The fit in this

case evaluates the compatibility of the given set of tracks with a b - c -hadron like cascade topology, increasing the discrimination power against light quark jets. The lateral displacement of the c -hadron decay vertex with respect to the b -hadron flight path is small enough not to violate significantly the basic assumption within the typical resolutions of the tracking detector. The discrimination between b -, c - and light jets is based on a likelihood using similar variables as in the SV1 tagging algorithm above, and additional variables such as the flight length significances of the vertices.

Algorithm combinations and performance

The IP3D and SV1 tagging algorithms both use the likelihood ratio method, and due to this they can be easily combined: the weights of the individual tagging algorithms are simply summed up.

The combination of the JetFitter and the IP3D algorithms can be performed using an artificial neural network technique with Monte Carlo simulated training samples and additional variables describing the topology of the decay chain.

Figure 4.4 compares the performance for the various ATLAS b -tagging algorithms described in a simulated sample of $t\bar{t}$ events. It can be seen that by combining the vertexing techniques and the impact parameter information, the IP3D+SV1 and IP3D+JetFitter algorithms can reach very high tagging efficiencies.

The performance of a b -tagging algorithm is usually measured in terms of the *light-jet rejection* obtained for a given *b -jet tagging efficiency*. Curves are obtained by varying continuously the *operating point* of each tagger, i.e. the cut on its output discriminating variable (weight). The b -jet tagging efficiency, ϵ_b , is the fraction of jets labeled as b -jets that are properly tagged

while the light-jet rejection, defined as $1/\epsilon_{light}$, is the reciprocal of the fraction of jets that are labeled as light jets and are actually tagged incorrectly by the algorithm.

The labeling procedure used for b -tagging is based on the flavor of true quarks: a jet is labeled as a b -quark jet if a b -quark is found in a cone of size $\Delta R = 0.3$ around the jet direction. The various labeling hypotheses are tried in this order: b quark, c quark and τ lepton. When none of these hypotheses are satisfied, the jet is labeled as a light jet. No attempt is made to distinguish light jets originating from gluons from those originating from quarks at this stage.

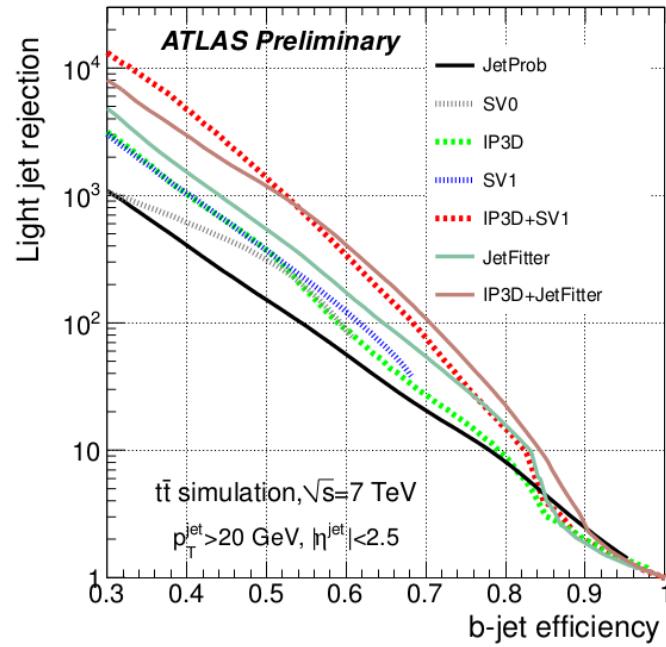


Figure 4.4: Light-jet rejection as a function of the b -jet tagging efficiency for the early tagging algorithms (JetProb and SV0) and for the high-performance algorithms, based on simulated $t\bar{t}$ events.

The MV1 tagging algorithm

The **MV1** b -tagging algorithm is a combined algorithm based on a neural network using the output weights of the IP3D and SV1 algorithms and the JetFitter+IP3D combination as input. For being the best performing algorithm (better light rejection for the same signal efficiency) it is the recommended tagger for 2011 and 2012 analyses. This is the b -tagging algorithm used in this thesis.

4.4.2 b -tagging calibration

In order for b -tagging to be used in physics analyses, the efficiency with which a jet originating from a b -quark is tagged by a b -tagging algorithm needs to be measured in data. Moreover, an appropriate description of the b -tagging efficiencies based on measurements with data is essential for correctly modelling the measurements in Monte Carlo simulation . A second necessary piece of information is the probability of mistakenly tagging a jet originating from a light-flavour (u -, d -, s -quark or gluon) jet as a b -jet, referred to as the mistag rate. The b -tagging “calibration” includes both the measurement of the mis-tag rates and b -tagging efficiency.

The measurements of the b -tag efficiency and mistag rate are provided in the form of jet p_T - and η -dependent scale factors that correct the b -tagging performance in simulation to that observed in data. The scale factors are defined as the ratio of the b -tag efficiency or mistag rate in data and simulation:

$$\kappa_{\epsilon_b}^{data/sim} = \frac{\epsilon_b^{data}}{\epsilon_b^{sim}}, \quad \kappa_{\epsilon_l}^{data/sim} = \frac{\epsilon_l^{data}}{\epsilon_l^{sim}}, \quad (4.13)$$

where ϵ_b^{sim} and ϵ_l^{sim} are the fractions of b - and light-flavour jets which are tagged in simulated events, with the jet flavour defined by matching to gen-

erator level partons as defined in the previous section.

In physics analyses, these p_T -dependent scale factors are then applied as weights to the jets in Monte Carlo simulation, to reproduce the b -tagging performance in data.

The main b -tagging efficiency calibration methods, the so called *system8* and *p_{rel}* methods are described in detail in ref [81]. These measurements are based on a sample of jets with muons inside, where the muons are serving as a reference b -tagging algorithm to obtain a b -jet sample on which the calibrations can be performed. At the LHC, the large $t\bar{t}$ production cross section of $\sigma_{t\bar{t}} = 1773(\text{stat.})+8(\text{sys.})7(\text{lum.}) \text{ pb}$ [82] offers an alternative source of events enriched in b -jets. Calibrations using samples of $t\bar{t}$ events have been obtained for SV0, IP3D+SV1, JetFitter and MV1 b -tagging algorithms [83]. All these algorithms provide an output weight w , discriminating between b -jets and non- b -jets. Lower values of w are assigned to c - and light-flavour jets, whereas the purity of b -jets increases with w . For each b -tagging algorithm a set of operating points, corresponding to a certain w cut value, are defined and calibrated:

- SV0: $\epsilon_b^{sim} = 50\%$
- IP3D+SV1: $\epsilon_b^{sim} = 60\%, \epsilon_b^{sim} = 70\%, \epsilon_b^{sim} = 80\%$
- JetFitter: $\epsilon_b^{sim} = 57\%, \epsilon_b^{sim} = 60\%, \epsilon_b^{sim} = 70\%, \epsilon_b^{sim} = 80\%$
- MV1: $\epsilon_b^{sim} = 60\%, \epsilon_b^{sim} = 70\%, \epsilon_b^{sim} = 75\%, \epsilon_b^{sim} = 85\%$

where ϵ_b^{sim} is the nominal b -tagging efficiency derived from an inclusive sample of simulated $t\bar{t}$ events.

The mistag rate is measured in data using two methods, both based on an inclusive sample of jets, referred to as the *negativetag* and *sv0mass*

methods [84]. The first method uses the invariant mass spectrum of tracks associated with reconstructed secondary vertices to separate light and heavy-flavour jets, and the other is based on the rate at which secondary vertices with negative decay length, or tracks with negative impact parameter, are present in the data.

Currently, there is no explicit measurement of the c -tag efficiency available in ATLAS. As both the b - and c -tag efficiencies are dominated by decays of long-lived heavy flavour hadrons, they are expected to show a similar behaviour. In general, for physics analyses, it is thus assumed that the scale factor is the same for b - and c -jets. However, to take into account possible deviations from this assumption, the systematic uncertainty for the c -tag efficiency scale factor is inflated by a factor of two, which is considered to be a conservative choice based on simulation studies. In the future, the c -tag efficiency will be measured in dedicated analyses.

Chapter 5

Double b -hadron jet identification

In this chapter we focus on the understanding of the internal structure of b -jets containing two b -hadrons by investigating the differences between these and single b -quark jets. These differences are expected to arise from the two-subjet structure of double b -hadron or “merged” jets, which would tend to be wider and with a larger number of constituents. Based on these envisaged characteristics, simulated QCD samples of b -tagged jets were used to explore properties with potential discrimination power. The Monte Carlo distributions were compared to data from the 2011 run for validation. We present results from these studies and discuss the choice of the observables selected to build the multivariable tool presented in Chapter 6.

5.1 Data Analysis

The tagging technique presented in this thesis relies on Monte Carlo predictions for the signal (single b) or background (merged b) hypotheses. The accuracy of the simulation is validated with data by comparing the distributions of the different variables studied.

The data samples employed correspond to proton-proton collisions at $\sqrt{s} = 7$ TeV delivered by the LHC and recorded by ATLAS between May and November 2011, with the LHC running with 50 ns bunch spacing, and bunches organized in bunch trains. Only data collected during stable beam periods in which all sub-detectors were fully operational are used. After the application of the data quality selection, the surviving data corresponds to an integrated luminosity of 4.7 fb^{-1} . The LHC instantaneous luminosity steadily increased during 2011. As a result, the average number of minimum-bias pile-up events, originating from collisions of additional protons in the same bunch as the signal collision, grew from 3 to 20. This fact will be of importance when discussing the selection of discriminating variables.

The Monte Carlo event generators discussed in Section 2.3.1 are used here. Samples of dijet events from proton-proton collision processes were simulated with PYTHIA version 6.423 [23], used both for the simulation of the hard $2 \rightarrow 2$ process as well as for the parton shower, underlying event, and hadronization models. The ATLAS AMBT2 tune of the soft model parameters was used [28]. In order to have sufficient statistics over the entire p_T spectrum, eight samples were generated with different thresholds of the hard-scattering partonic transverse momentum \hat{p}_T . Events from different samples were mixed taking into account their respective production cross sections. The simulated data sample used for the analysis gives an accurate description of the pile-up content and detector conditions for the full 2011 data-taking period.

5.1.1 Event selection

The event selection and quality criterion used to extract, from the data and Monte Carlo samples, the final set of jets for the analysis comprises different

steps:

- **Trigger.** The event sample was collected using the ATLAS single jet triggers which select events with at least one jet with transverse energy above a given threshold. At the hardware Level 1 and local software Level 2 (see Section 3.2.5), cluster-based jet triggers are used to select events with high- p_T jets. The Event Filter, in turn, runs the offline anti- k_t jet finding algorithm with $R = 0.4$ on topological clusters over the complete calorimeter. At this stage, the transverse energy thresholds, expressed in GeV, are: 20, 30, 40, 55, 75, 100, 135, 180. These triggers reach an efficiency of 99% for events having the leading jet with an offline energy higher than the corresponding trigger thresholds by a factor ranging between 1.5 and 2. The jet triggers with the lowest p_T thresholds were prescaled by up to five orders of magnitude.
- **Primary vertex.** The offline event selection requires at least one primary vertex candidate with 5 or more tracks. No requirements are placed on the longitudinal position (along the beam line) of the vertex as the beam spot is used as a constraint when fitting the vertex.
- **Primary jet algorithm.** The jet algorithm selected for the analysis was the ATLAS default anti- k_t algorithm [34], with a distance parameter $R = 0.4$, using calorimeter topological clusters [65] as input.
- **Jet calibration.** The EM+JES calibration scheme, described in Section 4.1, was used to correct the jet energies for inhomogeneities and for the non-compensating nature of the calorimeter.
- **Jet quality.** Several quality criteria are applied to eliminate “fake” jets that are caused by noise bursts in the calorimeters and energy

depositions belonging to a previous bunch crossing [85].

- **Jet tagging.** Only jets tagged as b -jets using the MV1 b -tagging algorithm at the 60% efficiency working point were considered.
- **Isolation.** b -tagged jets with close-by jets ($\Delta R < 0.8$) with p_T higher than 7 GeV at electromagnetic scale were not included in the analysis.

All jets, with transverse momentum between 40 and 480 GeV, the selected p_T range for the analysis, were required to be in a region with full tracking coverage, $|\eta_{jet}| < 2.1$, and they were classified in eight p_T bins chosen such as to match the jet trigger 99% efficiency thresholds (in GeV): 40, 60, 80, 110, 150, 200, 270, 360. An event is used if it satisfies the highest threshold trigger that is 99% efficient for the p_T bin that corresponds to the p_T of its leading jet.

In the case of MC, the reconstructed b -tagged jets were further classified into single and merged b -jets based on truth Monte Carlo information. A b -hadron is considered to be associated to a jet if the ΔR distance in $\eta - \phi$ space between the direction of the hadron and the jet axis is smaller than 0.4. Jets were labeled as merged (single) b -jets if they contained two (only one) b -hadron:

$$\text{single } b\text{-jets: } \Delta R(j, b_{1/2}) < 0.4 \quad (5.1)$$

$$\text{merged } b\text{-jets: } \Delta R(j, b_1) < 0.4 \& \Delta R(j, b_2) < 0.4 \quad (5.2)$$

where j is a jet in the event and $b_{1/2}$ are the b -hadrons in the event. In the case another size parameter is used for jet finding, the definitions in equations 5.1 and 5.2 change accordingly.

5.1.2 Track selection

It is important to select genuine tracks belonging to jets. Only tracks located within a cone of radius $\Delta R(j, \text{track}) \leq 0.4$ around the jet axis were considered. Cuts on $p_T^{\text{trk}} > 1.0$ GeV and the χ^2 of the track fit, $\chi^2/ndf < 3$, are applied. In addition, tracks are required to have a total of at least seven precision hits (pixel or micro-strip) in order to guarantee at least 3 z -measurements. Tracks are also required to fulfill cuts on the transverse and longitudinal impact parameters at the perigee to ensure that they arise from the primary vertex. As cutting on impact parameter (IP) significance might be detrimental for b -jets, where large IP values are expected, relaxed cuts were used, $|d_0| < 2$ mm, and $|z_0 \sin \theta| < 2$ mm, with θ being the polar angle measured with respect to the beam axis. The track quality cuts are summarized in table 5.1.

Track parameter	Selection
p_T	> 1 GeV
d_0^{PV}	< 2 mm
$z_0^{PV} \sin \theta$	< 2 mm
$\chi^2/ndof$	< 3
Number of Pixel hits	≥ 2
Number of SCT hits	≥ 4
Number of Pixel+SCT hits	≥ 7

Table 5.1: Track selection criteria used for double b -hadron jet tagging, where d_0^{PV} and z_0^{PV} denote the transverse and longitudinal impact parameters derived with respect to the primary vertex. The $\chi^2/ndof$ is that of the track fit.

5.2 Kinematic differences between single and double b -hadron jets

The differences between genuine b -quark jets and double b -hadron jets, that in QCD originate mainly from gluon splitting, are expected to arise from the two-subjet structure of merged jets. In this section we present the study of a set of jet shape and substructure variables for the discrimination between single and merged b -jets. These variables are built from jet constituents either at calorimeter level (topological clusters) or tracks associated to the jet.

Jet track multiplicity The jet track multiplicity is a variable simple to calculate that carries important information of the jet inner structure. It is defined as the number of tracks with p_T above 1 GeV, satisfying the quality cuts described in section 5.1.2, and contained within a cone of radius $R = 0.4$ around the jet axis. Figure 5.1 shows its distribution for two p_T bins, representative of the range covered in this study. It is observed that merged b -jets contain on average around two more tracks than single b -jets at low jet p_T , with a larger difference at higher p_T values.

The effect of the minimum track p_T requirement was examined by lowering the selection cut to $p_T > 0.5$ GeV. On the one hand this could lead to an improvement in discrimination if it captured more information about the fragmentation process; on the other hand, a lower minimum track p_T can make the method more sensitive to pile-up with the addition of soft tracks incorrectly associated to the jets. It was observed that reducing the p_T cut of the tracks degrades the discrimination because it widens the distributions without increasing the separation between single and merged jets.

In order to evaluate the effect of using displaced tracks originating from

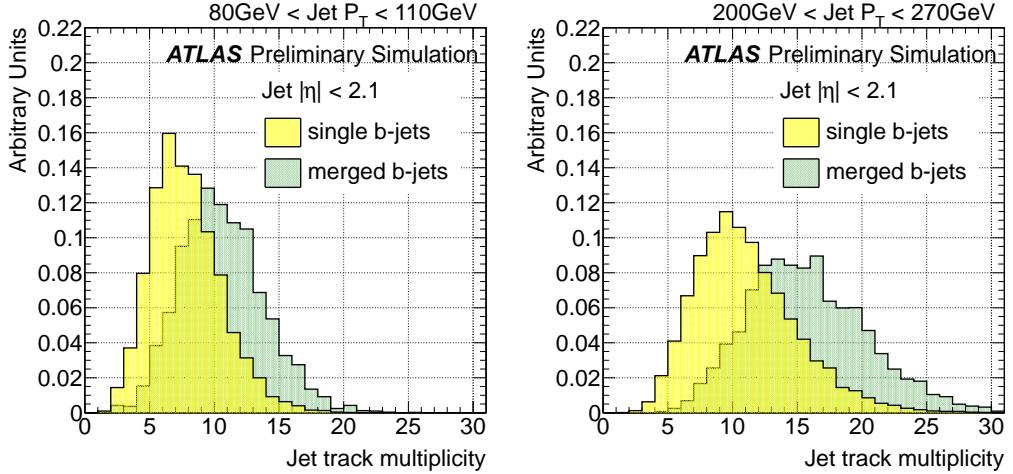


Figure 5.1: Distribution of the track multiplicity in jets for single and merged b -jets between 80 GeV to 110 GeV (left) and 200 GeV to 270 GeV (right).

the b -hadrons decays a preliminary study was done using a sample of dijet events generated with PYTHIA and with no detector simulation (standalone). For this study jets were reconstructed using all stable particles in the event, clustered with the anti- k_t algorithm. The association of charged particles was done in the same way as with the full ATLAS simulation. Distributions of the track multiplicity built using all charged particles and using only charged particles coming from the b -hadron decay (“ b -tracks”) are illustrated in Fig. 5.2. A better discrimination between single and merged b -jets is observed when using b -tracks only. The result obtained with PYTHIA standalone suggests that a potential improvement in single-merged separation can be achieved by circumscribing the track selection, in the full simulation, to tracks with large impact parameter significance. A comparison of track multiplicity distributions using all tracks and distributions built with displaced tracks is shown in Fig. 5.3. No improvement is obtained by using displaced tracks ($|d_0|/\sigma(d_0) > 2.5$). The potential sensitivity achieved by enriching the

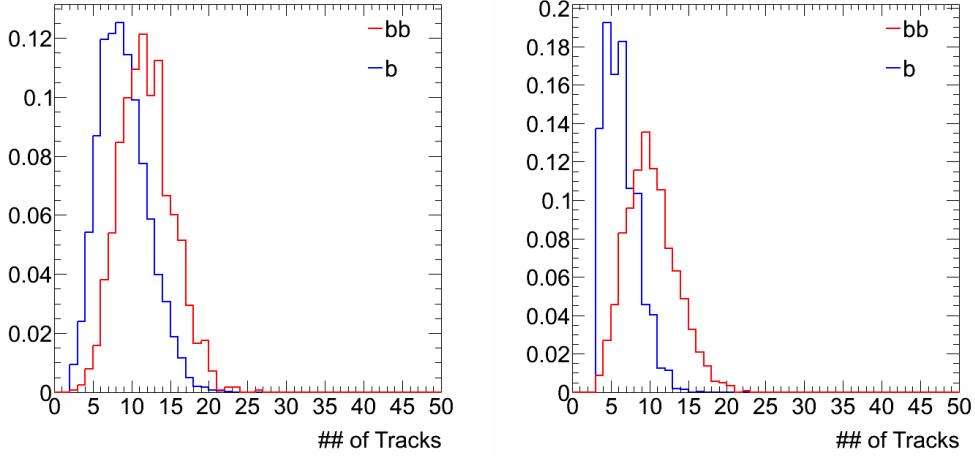


Figure 5.2: Distribution of the track multiplicity in jets for single (b) and merged (bb) jets between 80 GeV to 120 GeV in a sample of dijet events generated with PYTHIA and no detector simulation. Distributions are shown using all charged particles (left) and using only charged particles coming from b -hadron decay (right).

sample in tracks associated to the b -hadron is counterbalanced by the lower number of associated tracks.

Jet width The jet width is part of a set of continuous variables that attempt to distinguish individual particles/subjets within the jet by means of smooth function of $(\Delta\eta, \Delta\phi)$ away from the jet axis, in order to form combinations like geometric moments. This particular combination is a linear moment which sums the distances between the jet constituents and its axes, weighted by the constituent p_T , and normalized to the total p_T of the jet. Its definition is,

$$Jet\ width = \frac{\sum_{i=1}^N p_T^{const_i} \Delta R(const_i, jet)}{\sum_{i=1}^N p_T^{const_i}} \quad (5.3)$$

where N is the total number of calorimeter, track or particle constituents.

This observable has been extensively used to discriminate between gluon

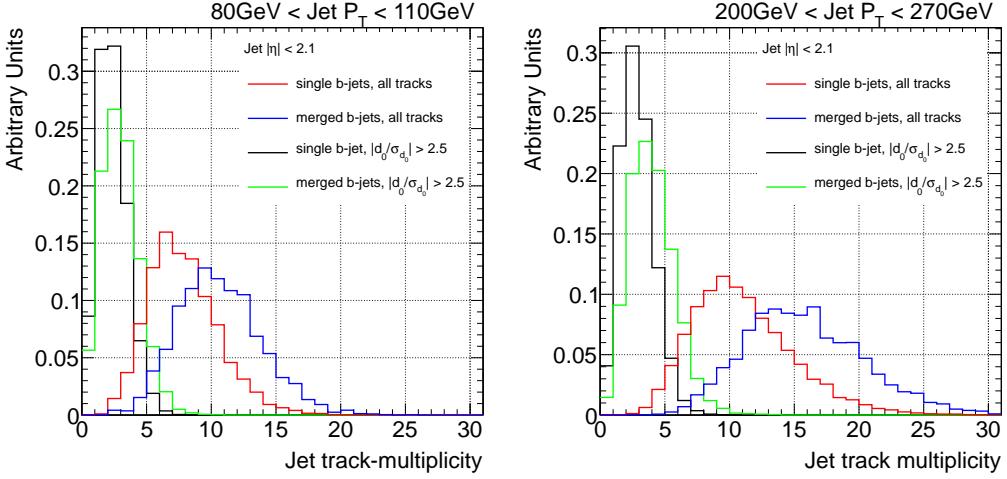


Figure 5.3: Distribution of the jet track multiplicity single and merged b -jets between 80 GeV to 110 GeV (left) and 200 GeV to 270 GeV (right), for all and displaced tracks only.

initiated and light quark initiated jets, see for instance [44] and [86]. Gluon jets are seen to be broader than quark jets. In the case of jets originating from b -quarks, these resemble gluon jets more closely than quarks jets [87]: due to the longer decay chain of b -hadrons the angular spread is larger for a b -jet than a light-quark jet.

In order to explore how merged jets, originating from a gluon splitting into a $b\bar{b}$ pair, compare to single b -quark jets and pure gluon jets, a standalone PYTHIA analysis was performed. Figure 5.4 illustrates the result; b -jets containing two b -hadrons present a greater angular width relative to single b -jets and gluon initiated jets. The latter, in turn, look broader than single for medium p_T jets. This behavior is somehow expected in the LHC's higher p_T jets because the QCD shower produces more particles resulting in broader gluon jets, whereas the particle multiplicity is relatively fixed in the b -hadron decay.

Distributions for the track-jet width with full ATLAS simulation are shown in Fig. 5.5. In this case the sum in equation 5.3 runs over the N tracks associated to the jet, using the same criteria as for the jet track multiplicity. As expected, merged b -jets are wider than single b -jets.

PYTHIA standalone samples were also used to evaluate the potential gain in discrimination obtained by utilising all stable particles in the event to build the observable, as opposed to using the charged particles only. A 10% improvement in merged b -jet rejection (for a 50% efficiency in selecting single b -jets) was achieved.

In full simulation, the jet width can be measured in terms of calorimeter variables by replacing tracks by topological clusters in the sum. Although it offers good separation, this variable is more sensitive to the amount of pile-up in the event than its track-based counterpart. This is illustrated in Fig. 5.6, which shows the distribution of calorimeter width and track-jet width for single b -jets in events with low and high number of primary vertices (NPV) in a low p_T region where the effect of pile-up is more important.

In general, all the studied calorimeter-based jet variables show similar dependences with NPV. For this reason the track-based versions are preferred as more robust discriminators.

Jet Mass The jet mass, like the linear radial moment, depends on the radiation pattern of the event. It is the most basic observable for distinguishing massive boosted objects from jets originating from quarks or gluons. The latter are expected to be dominated by wide-angle emissions, with increased probability to see high mass jets initiated from gluons as opposed to quarks [88].

Detector level jet mass distributions for jets selected to have $80 < p_T < 110$ GeV and $200 < p_T < 270$ GeV are shown in Fig. 5.7, both for single and

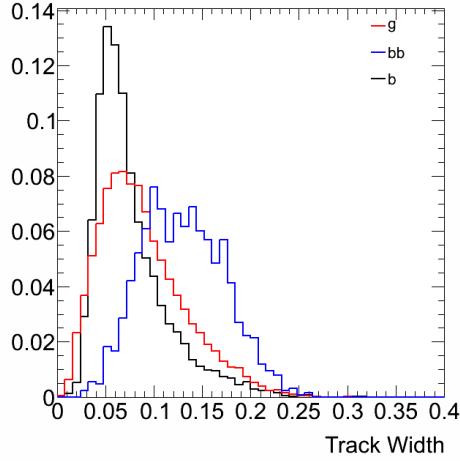


Figure 5.4: Distribution of track-jet width in jets for gluon-initiated (g), single (b) and merged (bb) jets between 80 GeV to 120 GeV in a sample of dijet events generated with PYTHIA and no detector simulation.

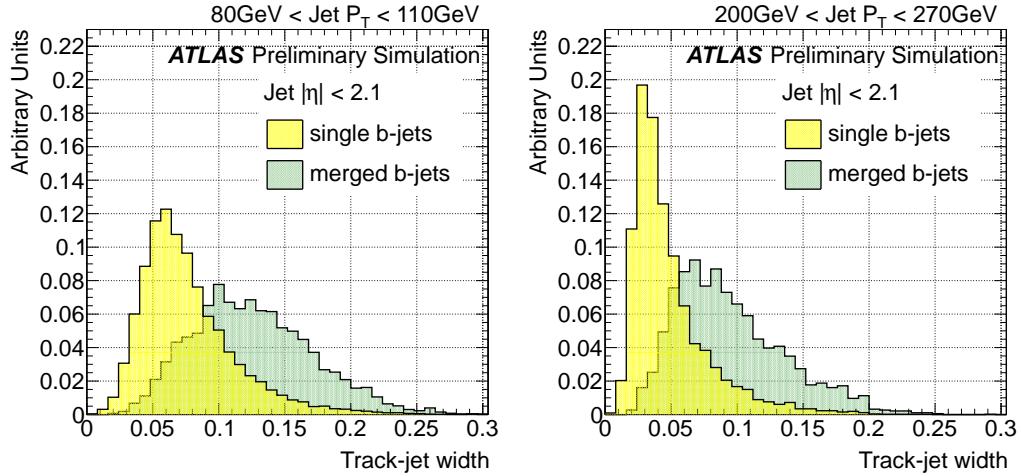


Figure 5.5: Distribution of track-jet width in jets for single and merged b -jets between 80 GeV to 110 GeV (left) and 200 GeV to 270 GeV (right).

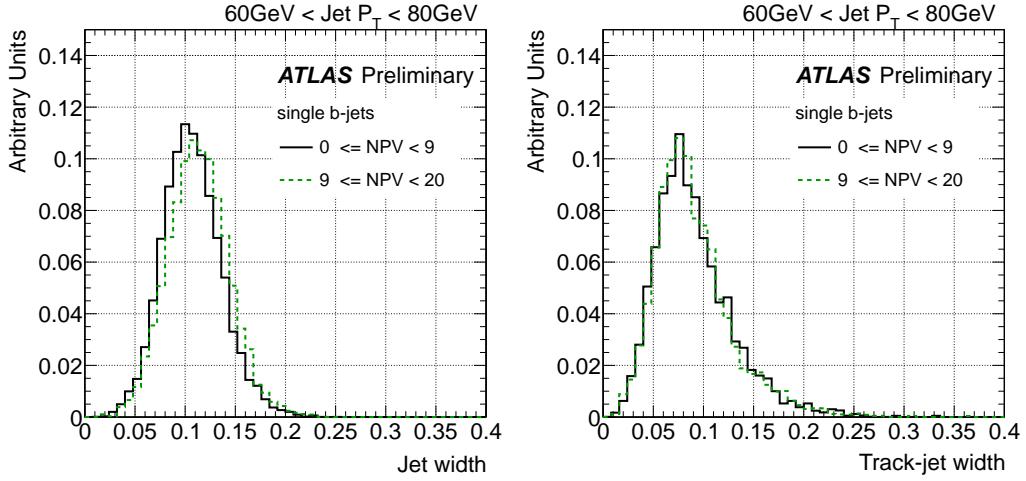


Figure 5.6: Distribution of jet width using topological clusters (left) and tracks (right) for single b -jets in two bins of number of primary vertices (NPV) for jets between 60 GeV to 80 GeV.

merged b -jets. Merged jets tend to have higher masses than single b -jets for the same p_T bin. Although it shows good separation, this calorimeter based variable can be significantly affected by the amount of pile-up in the event as even a single soft wide angle deposition will have an effect on the jet mass, shifting the distribution to higher values¹.

ΔR between leading tracks An alternative approach to measuring the width is to use the angular separation of the two hardest constituents inside jets. This has the advantage of removing any dependence on the shower development within the calorimeter and focuses on the hard components of the jet.

Figure 5.8 shows the distribution of the ΔR between leading tracks in the

¹In the ATLAS analysis of 35 pb^{-1} of data, the sensitivity of individual jet mass to pile-up is directly tested (for jets with at least 300 GeV). The mean jet mass is observed to increase linearly with NPV [89].

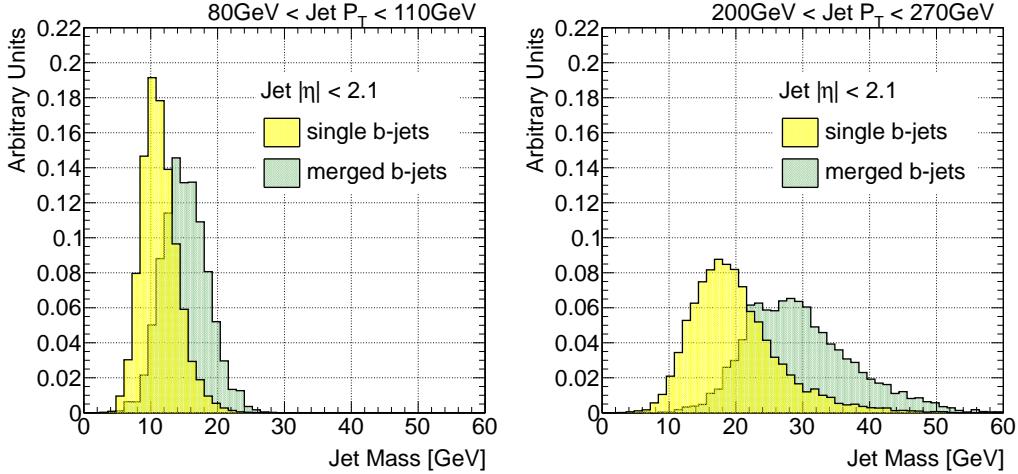


Figure 5.7: Distribution of jet mass in GeV for single and merged b -jets between 80 GeV to 110 GeV (left) and 200 GeV to 270 GeV (right).

jet for single and merged b -jets. The merged b -jet distributions are slightly broader than single b -jet distributions for medium jet p_T . The effect diminishes as we go to higher transverse momentum values, offering very poor discrimination.

Maximum ΔR between track pairs Several other variables, besides the jet width, were investigated to expose the expected two-subjet substructure of merged b -jets. The maximum ΔR separation between pairs of tracks associated to the jet ($\max\{\Delta R(trk, trk)\}$) was also evaluated as a discriminating variable. Its distribution is shown in Fig. 5.9, for single and double b -hadron jets. The latter show significantly higher values for this variable over a broad range of jet p_T . The distinct characteristic of this variable is that the separation between single b -jets and merged does not depend on jet p_T . In spite of its good discrimination power, alternative characterising variables are desirable as $\max\{\Delta R(trk, trk)\}$ is sensitive to soft tracks originating from pile-up.

Subjet multiplicity The subjet multiplicity – the number of subjets within

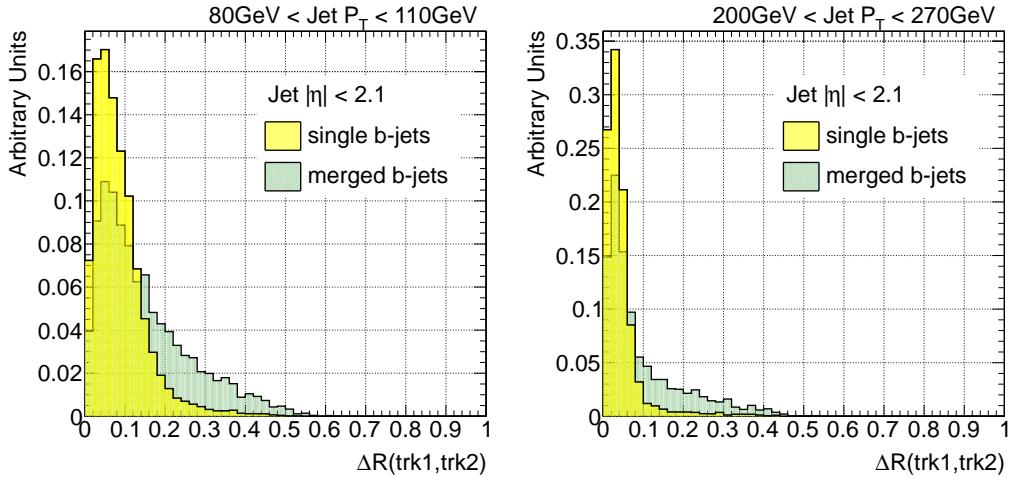


Figure 5.8: Distribution of ΔR between leading tracks for single and merged b -jets between 80 GeV to 110 GeV (left) and 200 GeV to 270 GeV (right).

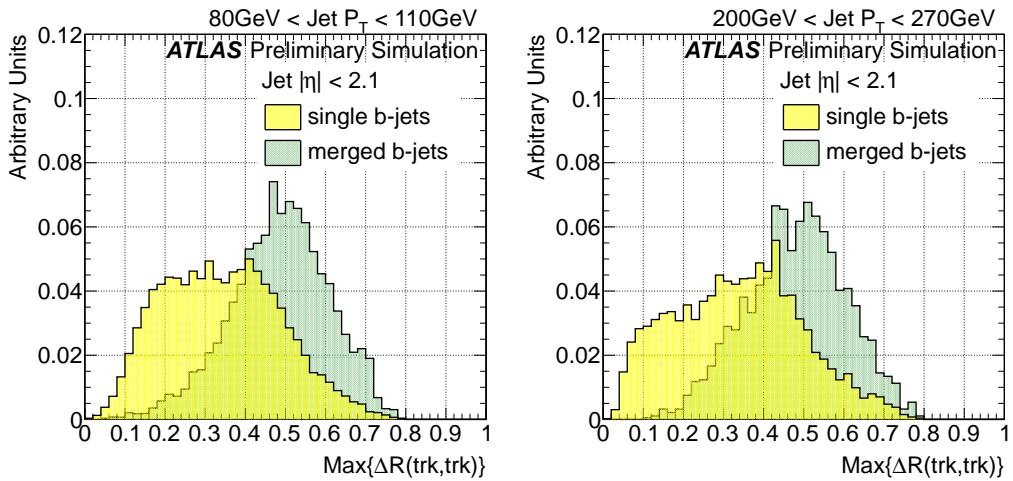


Figure 5.9: Distribution of the maximum ΔR between pairs of tracks in jets for single and merged b -jets between 80 GeV to 110 GeV (left) and 200 GeV to 270 GeV (right).

a jet – provides information on the distribution of energy and multiplicity of particles within a jet. For instance, a measurement of this observable for quark and gluon jets indicates that gluon-initiated jets tend to have on average higher subjet multiplicity [90]. This result is consistent with the QCD prediction that gluons radiate more than quarks. In this case the k_t algorithm is rerun for subjet finding.

In general, by using the inclusive k_t or other sequential recombination algorithms, introduced in Section 2.3.2, it is straightforward to define a “subjet algorithm” in which the structure of the jet constituents is resolved using either the same jet finder algorithm as used for jet reconstruction or a new one with a fixed (smaller) distance parameter. As an alternative to fixed distance parameter subjets, it is also possible to undo the last step in the recombination sequence [36] in order to identify the decay products of an object. This approach is used in several jet grooming procedures², see for instance [92].

Figure 5.10 shows the distribution of the number of subjets for single and merged b -jets. The subjets in this case were built using the associated tracks as constituents, clustered by the inclusive k_t algorithm with distance parameter $R = 0.2$. Merged jets tend to have on average one more subjet than single b -jets. The discrimination power of this variable is very poor.

ΔR between the axes of two k_t subjets The ΔR between k_t subjets is obtained by applying the exclusive k_t algorithm [93] to the tracks associated to the jet using a large k_t distance parameter to ensure that all tracks get combined. The clustering is stopped once it reaches exactly two jets. The ΔR between the axes of the two exclusive subjets is shown in Fig. 5.11. As

²Jet grooming comprises dedicated techniques to remove uncorrelated radiation within a jet. A review of these procedures can be found in [91].

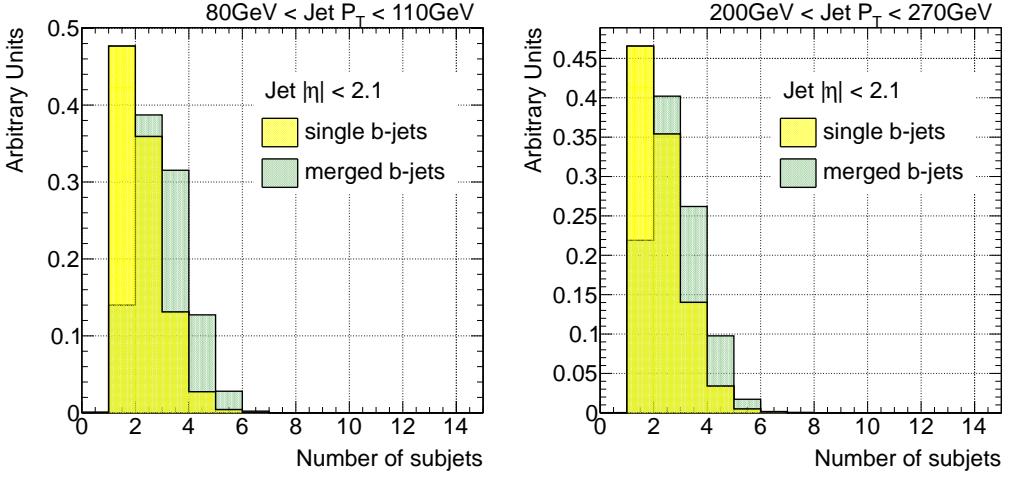


Figure 5.10: Distribution of the number of k_t sub-track-jets for single and merged b -jets between 80 GeV to 110 GeV (left) and 200 GeV to 270 GeV (right).

expected, it is larger for merged than for single jets. We observe that this variable provides very good separation, with the advantage of infrared safety and insensitivity to pile-up as opposed to $\max\{\Delta R((trk, trk)\}$.

In order to illustrate how this variable works, an event display of a merged b -jet with a large (> 0.3) ΔR value is shown in Fig. 5.12. For this image, the jet active area [94] was reconstructed following a procedure called “ghost association”, in which the event topoclusters plus a grid of 0.1×0.1 cells with nearly zero energy get associated to the jet through jet clustering. In the case of Figure 5.12 the anti- k_t algorithm was used. The event display indicates how the high energy cells in the jet with two b -hadrons are grouped around the b -hadrons directions, leading to the two-subjet substructure of merged jets.

N-subjettiness variables It is possible to extend the use of individual subjets in conjunction with more traditional jet shape variables. Using these

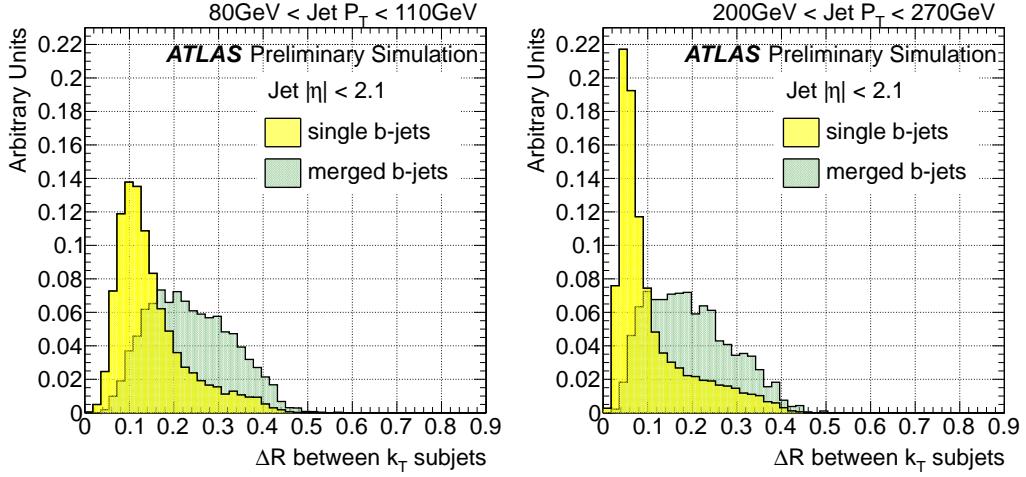


Figure 5.11: Distribution of the ΔR between the axes of the two k_t subjets in the jet for single and merged b -jets between 80 GeV to 110 GeV (left) and 200 GeV to 270 GeV (right).

tools, an inclusive jet shape based on the substructure topology of a single jet, “ N -subjettiness” [95] is defined. This variable describes the energy flow within a jet, quantifying the degree to which radiation is aligned along specified subjet axes. This jet shape was adapted from the event shape N -jettiness [96].

Given candidate subjets directions determined by an external algorithm such as the exclusive k_t procedure, the variable is defined as,

$$\tau_N^{(\beta)} = \frac{1}{\sum_k p_{Tk} (R_0)^\beta} \sum_k p_{Tk} (\min\{\Delta R_{j1,k}, \Delta R_{j2,k}, \dots, \Delta R_{jN,k}\})^\beta. \quad (5.4)$$

The sum runs over the k constituents in a given jet where p_{Tk} are their transverse momenta, and $\Delta R_{j1,k}$ is the distance between the candidate subjet $j1$ and a constituent particle k . R_0 is the characteristic jet radius used in the original jet clustering algorithm. The exponential weight, β , can optionally be applied to the angular distance computed between the subjets and the jet constituents. Since eq. 5.4 is linear in each of the constituent particle

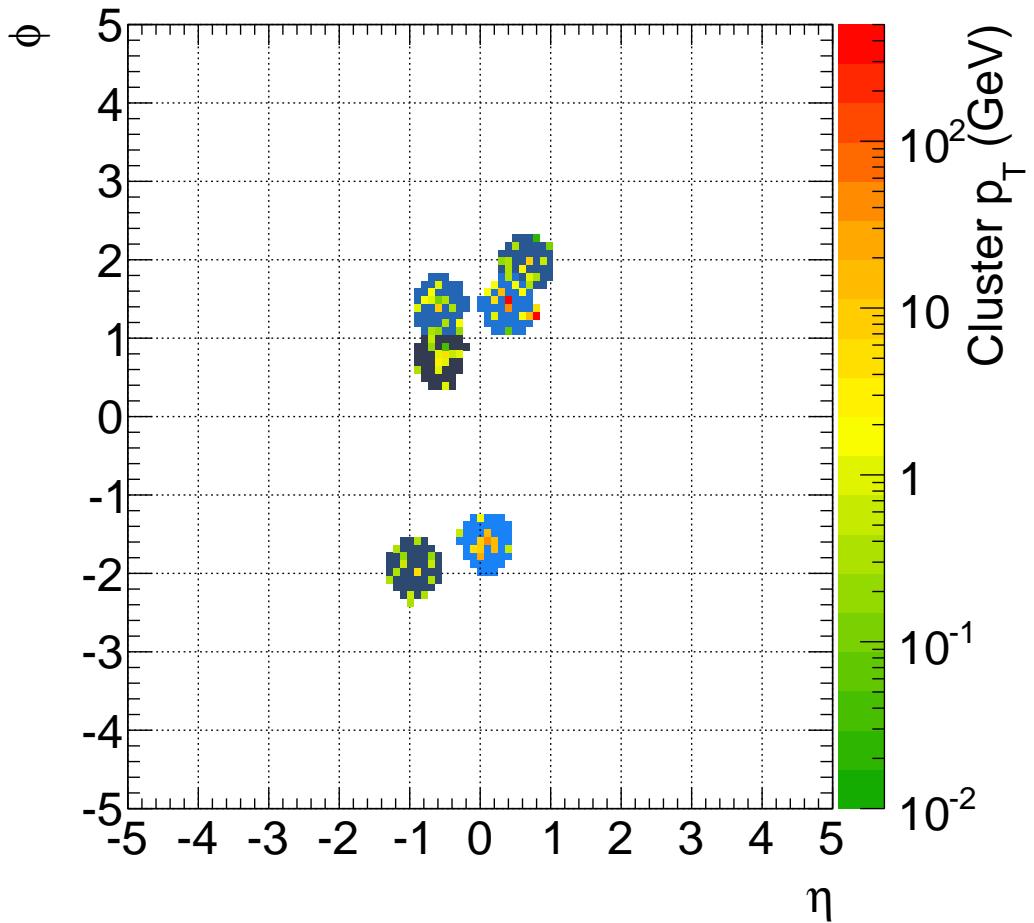


Figure 5.12: Event display of a merged b -jet in $(\eta, \phi) = (0.46, 1.41)$ and $p_T = 110$ GeV. The two b -hadrons are indicated as two red squares. The active area of the jet is shown in blue and the topoclusters belonging to the jet are shown in different colors, from green to orange, depending on their transverse momentum. The ΔR between the axes of the two k_t subjets in the jet is larger than 0.3. The two-subjet structure of the merged jet is displayed.

momenta, this variable is an infrared- and collinear-safe observable.

This jet shape was designed to separate boosted hadronic objects, like electroweak bosons and top quarks decaying into collimated showers of hadrons which a standard jet algorithm would reconstruct as single jets. A simple cut on the ratio τ_N/τ_{N-1} provides excellent discrimination power for N -prong hadronic objects[95]. In particular, τ_2/τ_1 can identify boosted W/Z and Higgs bosons, with the angular weighting exponent $\beta = 1$ providing the best discrimination.

The definition of N -subjettiness is not unique, and different choices can be used to give different weights to the emissions within a jet. The initial step of choosing candidate subjet axes is in fact unnecessary; the quantity in equation 5.4 can be minimised over the candidate subjet directions, further improving boosted object discrimination.

To avoid dependence on pile-up we consider track-based n -subjettiness, where the sum is over the tracks in the b -tagged jet. As seen for massive boosted objects, a jet with a two pronged structure, with all tracks clustered along two directions, is expected to have a smaller τ_2 value than a jet with a more uniform track distribution. The distributions of τ_2 , shown in Fig. 5.13, display good separation between single and merged jets, but with the latter showing larger values than single. This behavior can be traced to the level of correlation between τ_2 and track-jet width, displayed in Fig. 5.14a, to be compared to the much lower correlation presented, for instance, between track-jet width and the jet track multiplicity, shown in Fig. 5.14b.

The correlation observed suggests to switch from an absolute to a width-normalized τ_2 , and evaluate the ratio τ_2/τ_1 . Fig. 5.15 thus shows the distributions for this observable. Somewhat larger values are obtained for single than for merged b -jets, specially at high p_T , however we decided not to use

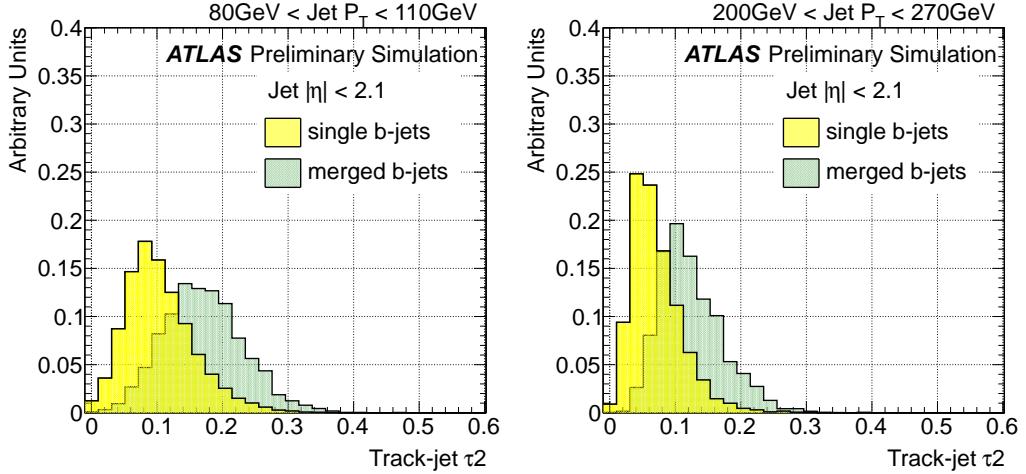


Figure 5.13: Distribution of τ_2 in jets for single and merged b -jets between 80 GeV to 110 GeV (left) and 200 GeV to 270 GeV (right).

this variable as it offers only marginal discrimination.

Jet eccentricity In defining a jet moment there are several ways to weight the momentum and define the center of the jet. We have defined the jet width as the first moment of the transverse energy with respect to the jet axis; another example of useful combination is the jet pull [97]. But it is also natural to look at higher moments, such as those contained in the 2×2 matrix,

$$\begin{bmatrix} \sum E_i \eta_i^2 & -\sum E_i \eta_i \phi_i \\ -\sum E_i \eta_i \phi_i & \sum E_i \phi_i^2 \end{bmatrix} \quad (5.5)$$

Here, (E_i, η_i, ϕ_i) are the jet constituent energy, pseudorapidity and azimuthal angle, respectively. The eigenvalues $\lambda_m \geq \lambda_p$ of this tensor are associated to the semiminor and semimajor axes of an elliptical jet, in the $\eta-\phi$ plane. The jet eccentricity, defined below, is a combination of these eigenvalues, and it is a measure of how elongated is the area of a jet,

$$e = \sqrt{1 - r^2} \quad (5.6)$$

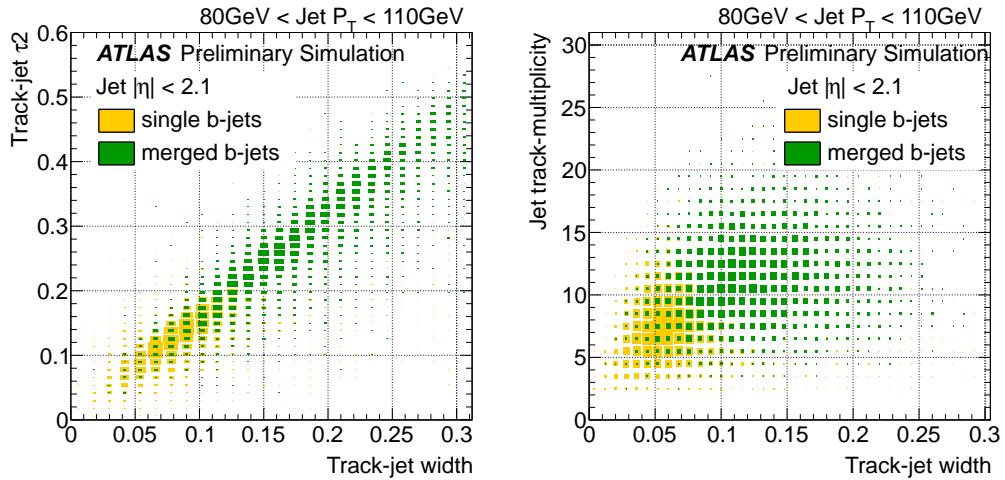


Figure 5.14: Correlation between τ_2 and track-jet width (left) and jet track multiplicity and track-jet width (right) for single and merged b -jets between 80 GeV to 110 GeV.

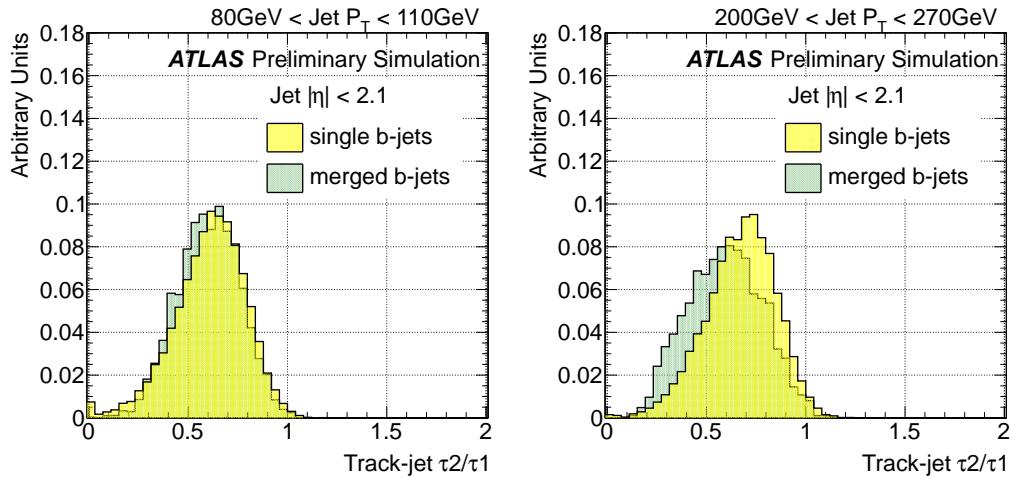


Figure 5.15: Distribution of τ_2/τ_1 in jets for single and merged b -jets between 80 GeV to 110 GeV (left) and 200 GeV to 270 GeV (right).

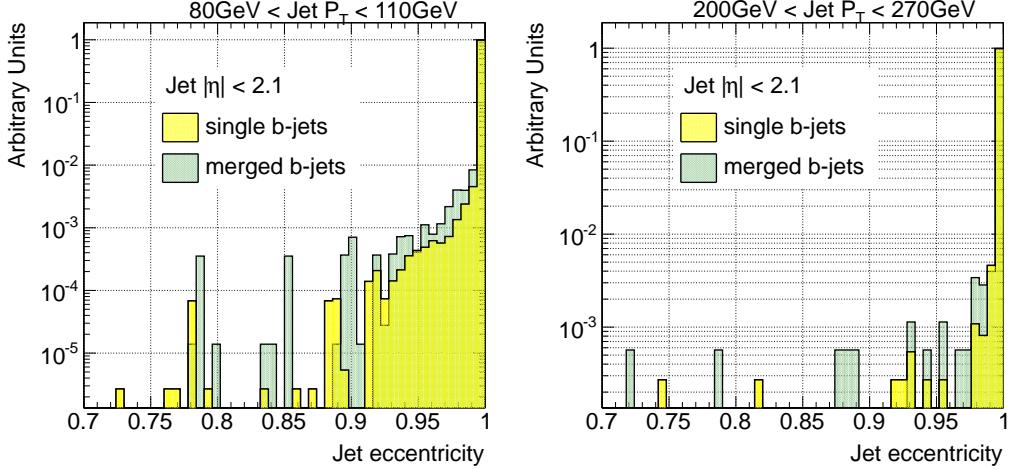


Figure 5.16: Distribution of the jet eccentricity for single and merged b -jets between 80 GeV to 110 GeV (left) and 200 GeV to 270 GeV (right).

where the parameter r is defined as the ratio of the eigenvalues,

$$r = \frac{\lambda_m}{\lambda_p} = \frac{\sum E_i \eta_i^2 + \sum E_i \phi_i^2 - \sqrt{(\sum E_i \eta_i^2 - \sum E_i \phi_i^2)^2 + 4(\sum E_i \eta_i \phi_i)^2}}{\sum E_i \eta_i^2 + \sum E_i \phi_i^2 + \sqrt{(\sum E_i \eta_i^2 - \sum E_i \phi_i^2)^2 + 4(\sum E_i \eta_i \phi_i)^2}}. \quad (5.7)$$

Figure 5.16 shows the distribution of the jet eccentricity, built using track constituents. No significant difference in eccentricity was found between single and merged b -jets.

In order to better understand the behavior observed for this variable, further studies were performed using bigger cone jets to enhance the efficiency to capture the decay products in gluon to $b\bar{b}$ -jets, in particular the anti- k_t 0.6 algorithm was used; and the ghost association procedure for cluster matching, for a calorimeter-based eccentricity. None of the above resulted in better single-merged separation.

5.3 Validation of the jet variables in data

In order to study the extent to which the simulation reproduces the distributions observed in data for the different variables explored a set of comparison plots is presented. Figures 5.17 and 5.18 show distributions of jet track multiplicity, track-jet width, ΔR between the axes of the two k_t subjets, $\max\{\Delta R(trk, trk)\}$ and τ_2 in two different p_T bins for b -tagged jets in di-jet Monte Carlo and data events passing selection described in Section 5.1.1. The distributions are normalized to unit area to allow for shape comparisons. There is a good agreement between data and simulation.

It should be remarked that the observed agreement is actually not a direct validation of the description in the MC of the relevant variables, but its convolution with the simulated relative fractions of light-, c -, b - and bb -jets in the b -tagged generated jet sample. To some extent, there could be some level of compensation between these two effects, although the agreement evaluated in b -jets selected with a looser cut of MV1 tagger as well as with another b -tagging algorithm is still very good, suggesting that this compensation is not likely to occur in samples sufficiently enriched in b -jets.

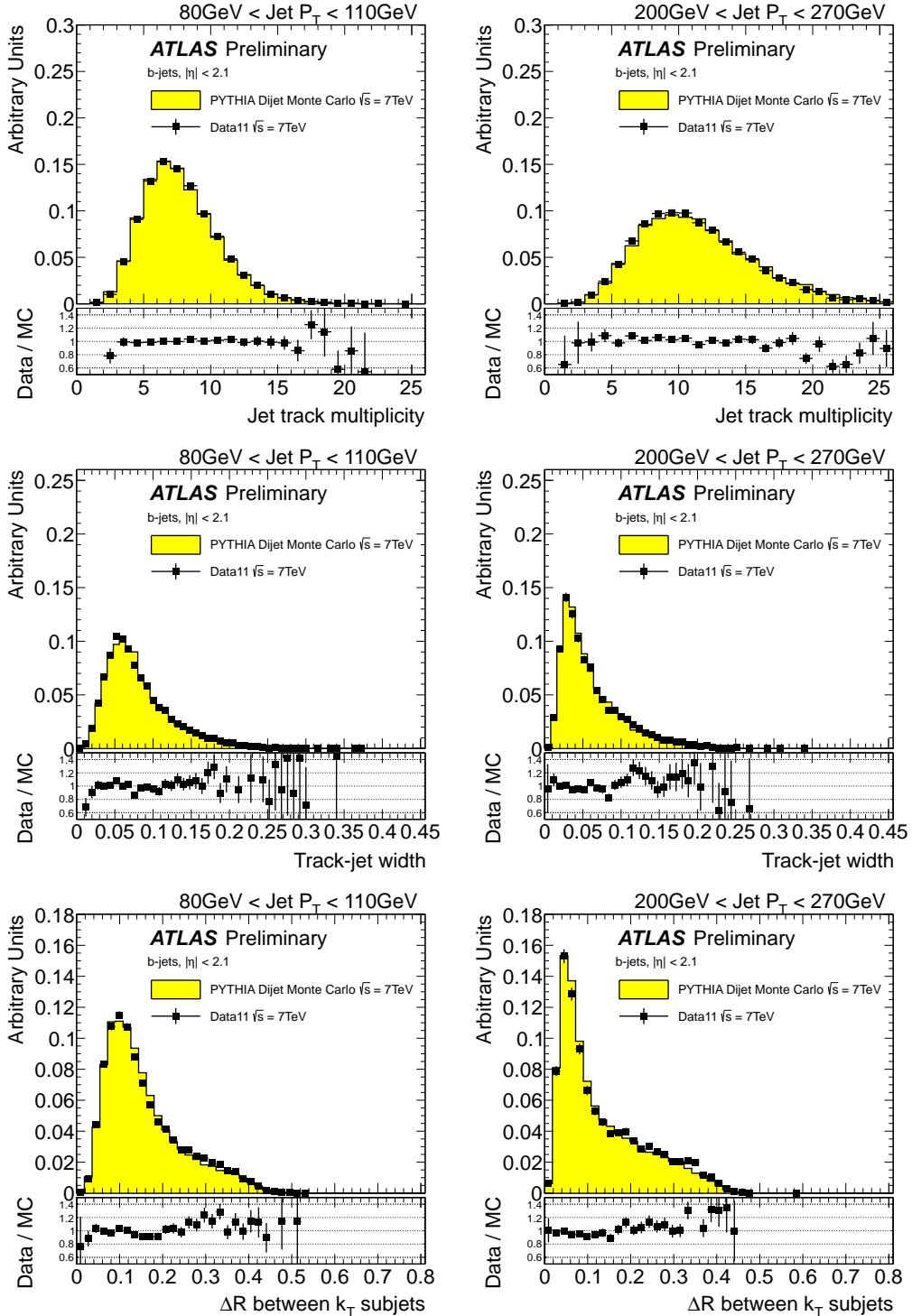


Figure 5.17: Distribution of three tracking variables in 2 different jet p_T bins, for experimental data collected by ATLAS during 2011 (solid black points), and simulated data (filled histograms). The ratio data over simulation is shown at the bottom of each plot.

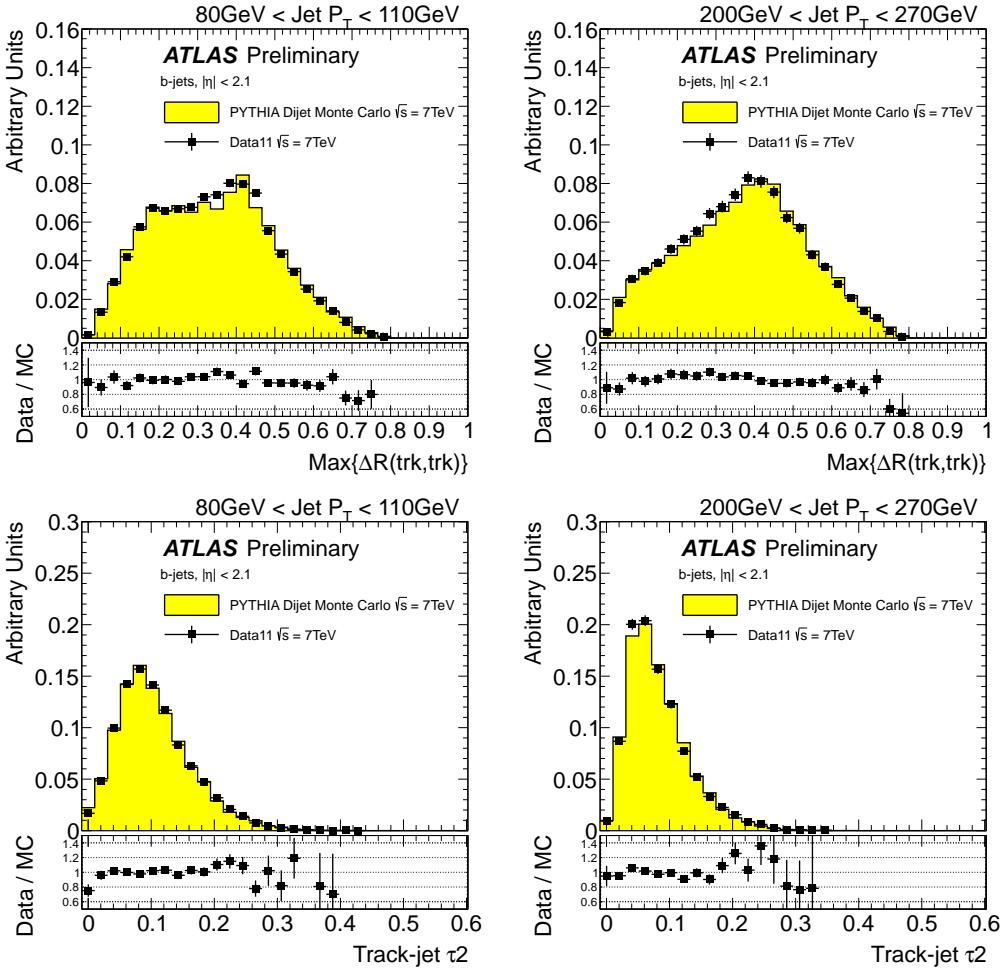


Figure 5.18: Distribution of two tracking variables in two different jet p_T bins, for experimental data collected by ATLAS during 2011 (solid black points), and simulated data (filled histograms). The ratio data over simulation is shown at the bottom of each plot.

Chapter 6

The Multivariate Analysis

After the evaluation of the best discriminating variables, a tagging algorithm, capable of efficiently identifying single b -jets while rejecting merged b -jets, can be constructed using several different approaches. This analysis exploits a multivariate likelihood approach for the separation of the single b -jet signal and its background (the merged b -jets). Compare to single variable or 2-dimensional cuts analyses, the sensitivity can be largely improved because several variables are combined to build a likelihood classifier with high separation power. Variables considered for the likelihood discriminant include the charged particle multiplicity as well as variables that account for the angular spread of the b -jets considered.

6.1 Multivariate methods

Multivariate data analysis refers to a statistical technique used to analyze data that arises from more than one variable. Classification is done through learning algorithms that make use of training events, for which the desired output is known, to determine the mapping function that describes a decision

boundary. The following multivariate methods were explored:

- Likelihood ratio estimators
- Neural Networks (NN)
- Boosted decision Trees (BDTs)

The method of maximum likelihood consists of building a model out of probability density functions (PDF) that reproduces the input variables for signal and background. For a given event, the likelihood for being of signal type is obtained by multiplying the signal probability densities of all input variables and normalising this by the sum of the signal and background likelihoods. The likelihood ratio $y_L(i)$ for event i is defined by:

$$y_L(i) = \frac{L_S(i)}{L_S(i) + L_B(i)}, \quad (6.1)$$

where

$$L_{S(B)}(i) = \prod_{k=1}^{n_{var}} p_{S(B),k}(x_k(i)), \quad (6.2)$$

and where $p_{S(B),k}(x_k(i))$ is the signal (background) PDF for the k th input variable x_k . All the PDFs are normalized to one.

The parametric form of the PDFs is generally unknown, however it is possible to empirically approximate its shape by nonparametric functions. Nonparametric models differ from parametric models in that the model structure is not specified a priori but is instead determined from data sample used for training. A histogram is a simple example of a nonparametric estimate of a probability distribution. The nonparametric functions can be chosen individually for each variable and can be either polynomial splines of various degrees fitted to binned histograms¹ or unbinned kernel density estimators

¹A spline is a sufficiently smooth polynomial function that is piecewise-defined, and possesses a high degree of smoothness at the places where the polynomial pieces connect. It is often referred to as polynomial interpolation.

(KDE). The idea behind the latter approach is to estimate the shape of a PDF by the sum over smeared training events. For a PDF $p(x)$ of a variable x , one finds [98].

$$p(x) = \frac{1}{Nh} \sum_{i=1}^N K\left(\frac{x - x_i}{h}\right) = \frac{1}{N} \sum_{i=1}^N K_h(x - x_i), \quad (6.3)$$

where N is the number of training events, $K_h(t) = K(t/h)h$ is the kernel function, and h is the bandwidth of the kernel (also termed the *smoothing parameter*). For the present implementation a Gaussian form of K is used.

The smoothness of the kernel density estimate is evident compared to the discreteness of a histogram; kernel density estimates converge faster to the true underlying density for continuous random variables.

An artificial Neural Network (NN) is a nonlinear discriminant. It is, most generally speaking, any simulated collection of interconnected neurons, with each neuron producing a certain response at a given set of input signals. It can be viewed as a mapping from a space of input variables $x_1, \dots, x_{n_{var}}$ onto, in the case of a signal-versus-background discrimination problem, a one-dimensional output variable. The behaviour of an artificial neural network is determined by the layout of the neurons, the weights of the inter-neuron connections, and by the response of the neurons to the input, described by the neuron response function. The neuron response function maps the neuron input (in R^n) onto the neuron output (R); often it can be separated into a synapse function ($R^n \rightarrow R$) and a neuron activation function ($R \rightarrow R$). The neuron activation function can be either a *linear*, *sigmoid*, *tanh*, or a *radial* function.

While in principle a neural network with n neurons can have n^2 directional connections, the complexity can be reduced by organising the neurons in layers and only allowing direct connections from a given layer to the following

layer. This kind of neural network is termed multi-layer perceptron. The first layer of a multilayer perceptron is the input layer, the last one the output layer, and all others are hidden layers. For a classification problem with n_{var} input variables the input layer consists of n_{var} neurons that hold the input values, $x_1, \dots, x_{n_{var}}$, and one neuron in the output layer that holds the output variable, the neural net estimator y_{NN} .

Decision trees are well known classifiers that allow a straightforward interpretation as they can be visualized by a simple two-dimensional tree structure. They are in this respect similar to rectangular cuts. Repeated yes/no decisions are taken on one single variable at a time until a stop criterion is fulfilled. The phase space is split this way into many regions that are eventually identified as “signal-like” or “background-like”, depending on the majority of training events that end up in the final *leaf* node. The boosting of a decision tree extends this concept from one tree to several trees which form a *forest*. Boosting increases the statistical stability of the classifier and typically also improves the separation performance compared to a single decision tree [99].

Training and testing the MVA methods

A sub-set of the dijet Monte Carlo sample was used for training the three methods in the context of the Toolkit for Multivariate Data Analysis, TMVA [100], written in C++ language. After the event and jet selections were performed, the b -tagged jets with $|\eta| < 2.1$ were classified as signal (single b -jets) or background (merged b).

Based on discrimination power, correlation and pile-up dependence three variables were selected for the training: the jet track multiplicity, the track-jet width and the ΔR between the axes of two k_t subjets in the jet. Distributions

of the these variables for single and merged b -jets were given to the multivariate methods as input. Other variables such as τ_2 or $\max\{\Delta R(trk, trk)\}$ were also tested leading to no gain in performance.

Different training options were evaluated for the likelihood and Neural Network classifiers. The final configuration for the likelihood estimator uses the KDE approach for PDF estimation of the input variables. The NN trained is a Multi-layer perceptron (MLP) with two hidden layer of n_{var} and $n_{var} - 1$ neurons respectively, using a sigmoidal neuron activation function. The Boosted Decision Tree approach was implemented with 400 trees in the BDT forest (double the TMVA default value). Distributions of the final output for the three methods, evaluated in a sample of simulated dijet events, are shown in Fig. 6.2 for a medium p_T bin. The testing sample satisfies the same selection used for the training sample.

The outputs of the MVA discriminants explored are different in terms of shape and range, although the latter could be rearrange with a suitable variable transformation. In spite of these distinct features the performances of the different methods, expressed in rejection of merged b -jets as a function of single b -jet efficiency (see next section), agrees within statistics, see Fig. 6.2.

As opposed to Neural Network discriminants with large number of training cycles, the training and the application of the likelihood are very fast operations that are suitable for very large data sets and tuning of the training parameters. Although also very fast, a shortcoming of decision trees is their instability with respect to statistical fluctuations in the training sample from which the tree structure is derived. If two input variables exhibit similar separation power, a fluctuation in the training sample may cause the tree growing algorithm to decide to split on one variable, while the other variable could have been selected without that fluctuation. In such a case

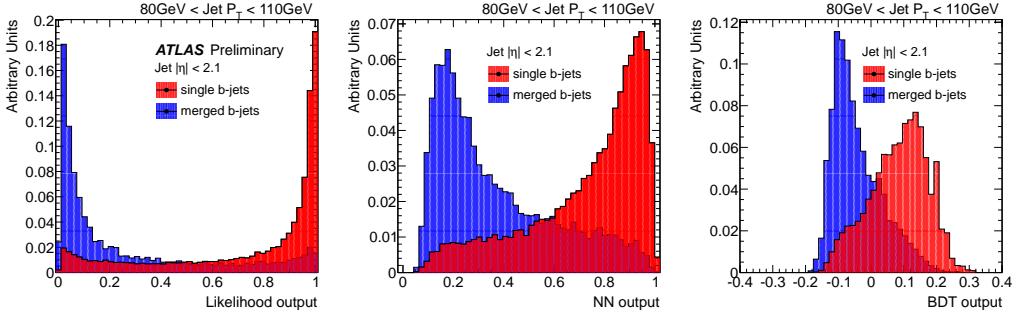


Figure 6.1: Distribution of the MVA discriminant outputs for the Likelihood (a), Neural Network (b) and Boosted Decision Trees (c) classifiers, for single and merged b -jets between 80 GeV and 110 GeV.

the whole tree structure is altered below this node, possibly resulting also in a substantially different classifier response [100]. Is for these reasons that the likelihood classifier is the selected method for our tagger.

6.2 Likelihood training and performance

As indicated in section 6.1, a discriminant between single and merged b -jets was built by training a simple likelihood estimator, with the following combination of three variables as input,

1. Jet track multiplicity
2. Track-jet width
3. ΔR between the axes of 2 k_t subjets within the jet

Given the correlation of the variables with the jet transverse momentum, the training sample was categorized in bins of calorimeter jet p_T , and independent likelihood classifiers were built for each category. Signal and background jets were not weighted by the dijet samples cross-sections to allow the contribution of subleading lower p_T jets from high p_T events, and thus

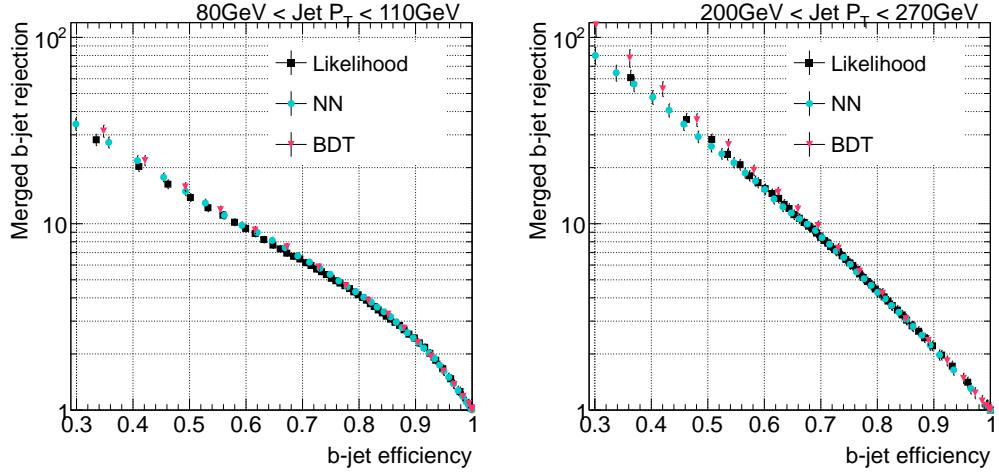


Figure 6.2: Rejection of merged b -jets as a function of single b -jet efficiency for the the different MVA methods evaluated for low and high jet p_T .

increase the statistics of merged jets in the low p_T bins. For the evaluation of the method the same procedure was followed.

The distribution of the likelihood output for single and merged b -jets is shown in Fig. 6.3 for low, medium and high transverse momentum jets.

The performance of the tagger in the simulation can be displayed in a plot of rejection ($1/\epsilon_{bkg}$) of merged b -jets as a function of single b -jet efficiency, where ϵ_{bkg} is the probability that a double b -hadron jet passes the single b -jet tagger. This is shown in Fig. 6.4 for the eight bins of jet p_T mentioned in section 5.1.1. The performance improves with p_T :

- $p_T > 40$ GeV: rejection above 8 at 50% eff.
- $p_T > 60$ GeV: rejection above 10 at 50% eff.
- $p_T > 200$ GeV: rejection above 30 at 50% eff.

The rejection of merged jets attained as a function of p_T for the 50% and 60% single b -jet efficiency working points are summarized in Table 6.1,

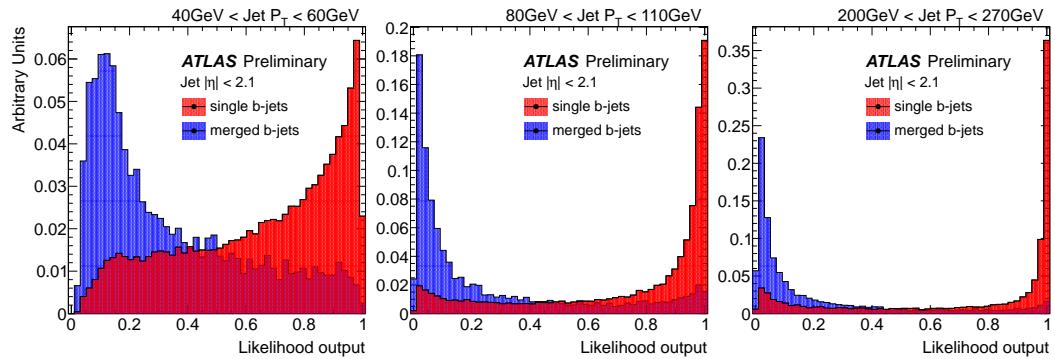


Figure 6.3: Distribution of the likelihood output for single and merged b -jets for low, medium and high p_T jets.

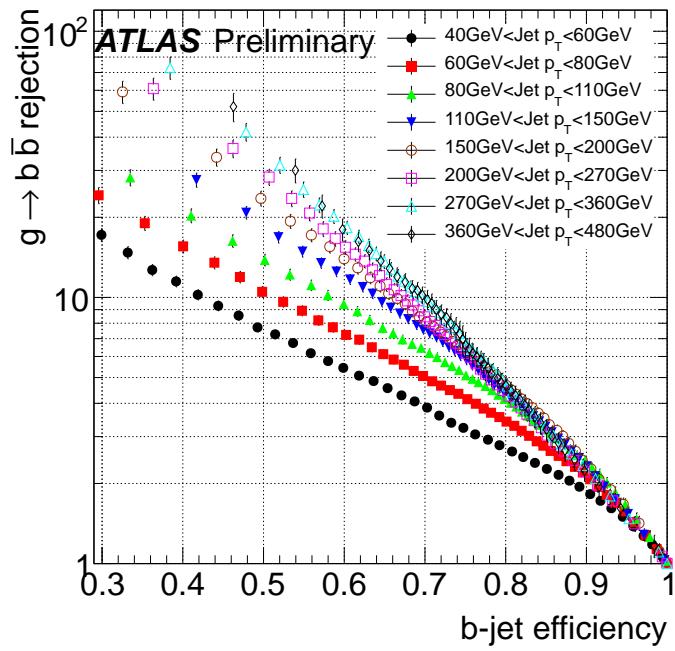


Figure 6.4: Rejection of merged b -jets as a function of single b -jet efficiency for dijet events in 8 jet p_T bins.

together with their relative statistical error. These are propagated from the Poisson fluctuations of the number of events in the merged and single b distributions. The error is slightly lower for the 60% efficiency working point because a higher efficiency allows for a greater number of Monte Carlo events to measure the performance.

Jet p_T (GeV)	single b -jet efficiency 50%		single b -jet efficiency 60%	
	Rejection	stat.err.	Rejection	stat.err.
40 - 60	8	4%	5	3%
60 - 80	10	4%	7	4%
80 - 110	14	5%	9	4%
110 - 150	19	5%	12	4%
150 - 200	23	5%	14	5%
200 - 270	30	7%	16	6%
270 - 360	36	7%	19	6%
360 - 480	41	8%	18	8%

Table 6.1: The merged b -jet rejection for the 50% and 60% efficiency working points in bins of p_T .

6.3 Systematic uncertainties

The development, training and performance determination of the tagger is based on simulated events. Although the agreement between simulation and data explored in section 5.3 is a necessary validation condition, it is also important to investigate how the tagger performance depends on systematics

relevant in the data. In particular we have considered:

- presence of additional interactions (pile-up);
- uncertainty in the b -jet tagging efficiency;
- uncertainty in the track reconstruction efficiency;
- uncertainty in the track transverse momentum resolution;
- uncertainty in the jet transverse momentum resolution;
- uncertainty in the b -jet energy scale.

I. Pile-up

The size of this effect was studied by comparing the performance of the likelihood discriminant with b -jets in events with small (1-9) and large (9-20) number of primary vertices. A comparison of the performance in these two sub-samples relative to the inclusive sample is shown in Fig. 6.5. As expected from the use of tracking (as opposed to calorimeter) variables no significant dependence with pile-up is observed within statistics. Performance differences between high and low number of primary vertices events are $\leq 0.3\%$ and therefore negligible compared to other sources of uncertainties. The impact of pile-up might be larger in 2012.

II. b -tagging efficiency

The performance of heavy-flavor tagging in Monte Carlo events is calibrated to experimental data by means of the scale factors (SFs). The SFs are defined as the ratio of the heavy-flavor tagging efficiency in data over that in Monte Carlo for the different jet flavors. They are measured by the ATLAS Flavour Tagging Working group, and their measurement carries a systematic uncertainty.

To estimate the impact of this uncertainty a conservative approach is followed: the SFs are varied in all the p_T bins simultaneously by one standard

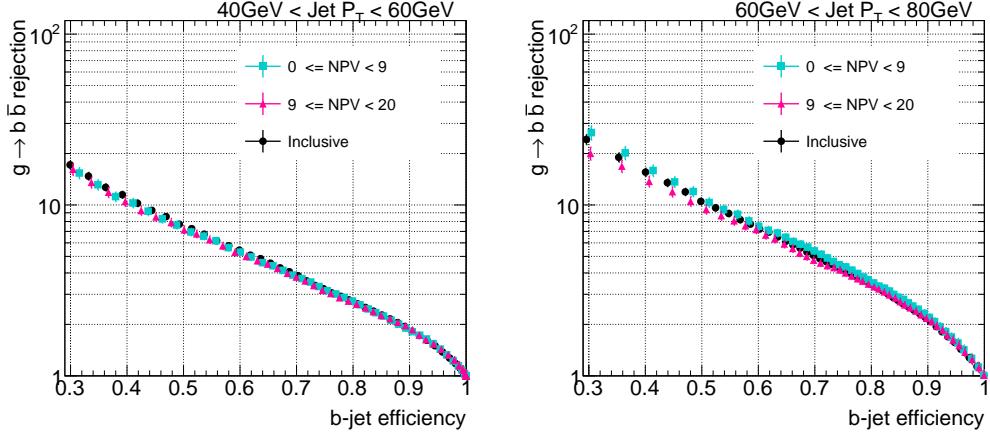


Figure 6.5: Rejection of merged b -jets as a function of single b -jet efficiency in bins of NPV for two low jet p_T bins.

deviation both in the up and down directions. The MC distributions weighted by the varied SFs show no major deviations from the nominal, see Fig. 6.6. In the same manner, the effect of the b -tagging calibration uncertainty on the likelihood performance, shown in Fig. 6.7, is $< 1\%$, negligible with respect to the statistical uncertainty. This was indeed expected. The scale factors depend on the true flavor of the jet and on its p_T , but these are basically constant in the performance determination, which is based on single flavor (true b -) jets classified in p_T -bins.

III. Track reconstruction efficiency

This uncertainty arises from the limit in the understanding of the material layout of the Inner Detector. To test its impact a fraction of tracks determined from the track efficiency uncertainty was randomly removed.

The tracking efficiency systematics are given in bins of track η . For tracks with $p_T^{\text{track}} > 500$ MeV the uncertainties are independent of p_T : 2% for $|\eta^{\text{track}}| < 1.3$, 3% for $1.3 < |\eta^{\text{track}}| < 1.9$, 4% for $1.9 < |\eta^{\text{track}}| < 2.1$, 4%

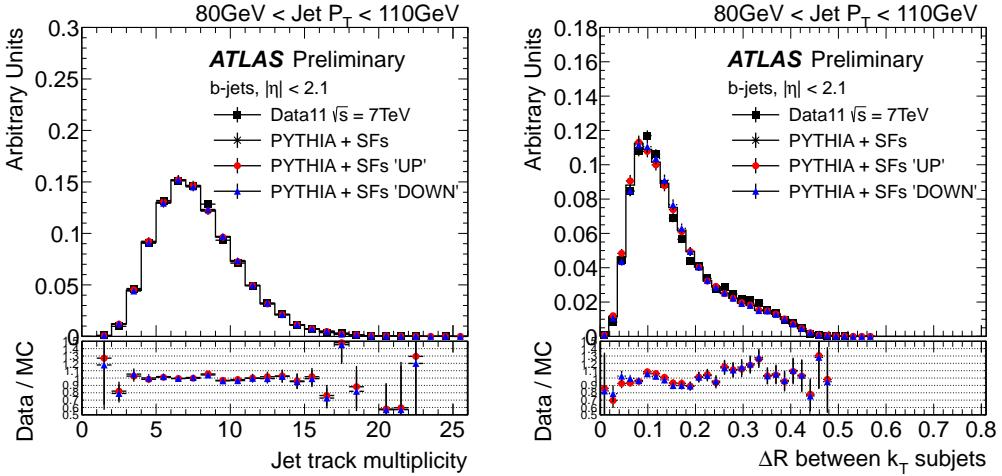


Figure 6.6: The effect of a variation in the b -tagging Scale Factors on the tracking variables distributions. Scale Factors were varied up (down) by 1-sigma to evaluate the systematic uncertainty from this source. The ratio data over MC is shown for MC PYTHIA with SFs varied up (circles) and down (triangles).

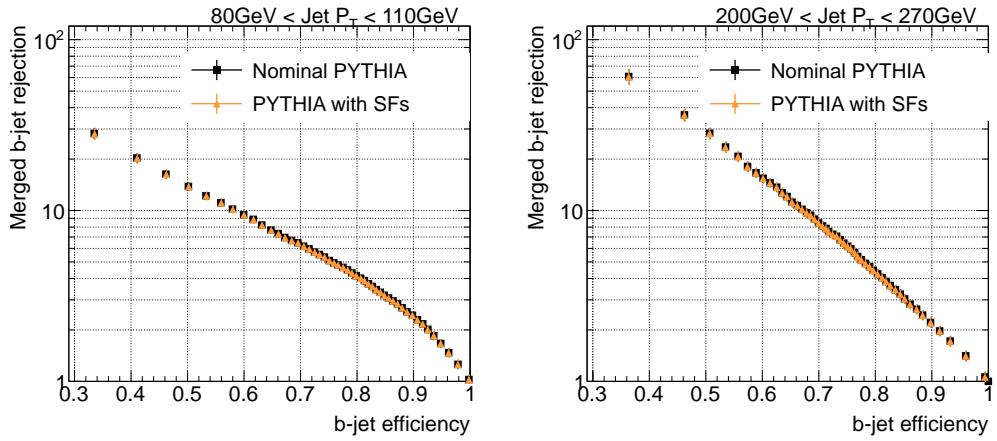


Figure 6.7: Rejection of merged b -jets as a function of single b -jet efficiency with and without scale factors as weights.

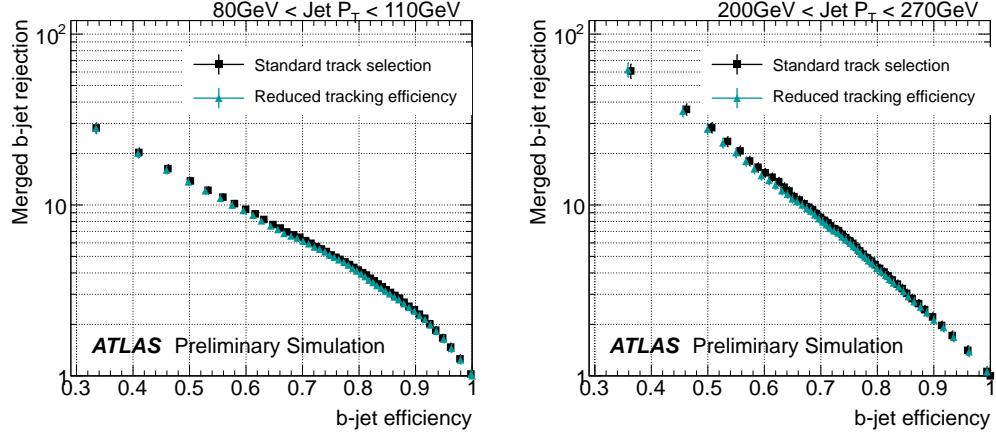


Figure 6.8: Rejection of merged b -jets as a function of the single b -jet efficiency showing shift in likelihood performance caused by a reduction in the tracking efficiency.

for $2.1 < |\eta^{\text{track}}| < 2.3$ and 7% for $2.3 < |\eta^{\text{track}}| < 2.5$ [59]. All numbers are relative to the corresponding tracking efficiencies.

The tracking variables were re-calculated and the performance of the nominal likelihood was evaluated in the new sample with worse tracking efficiency. The rejection-efficiency curves show a small degradation of the performance which is comparable to the statistical uncertainty. The effect is however systematically present over all 16 p_T bin/working points, without a clear p_T dependence. We have thus taken the average over p_T , and obtained a global systematic uncertainty of 4% both for the 50% and 60% efficiency working points. The performance comparison is shown in Fig. 6.8 for two p_T bins.

IV. Track momentum resolution

The knowledge of the track momentum resolution is limited by the precision both in the material description of the Inner Detector and in the mapping

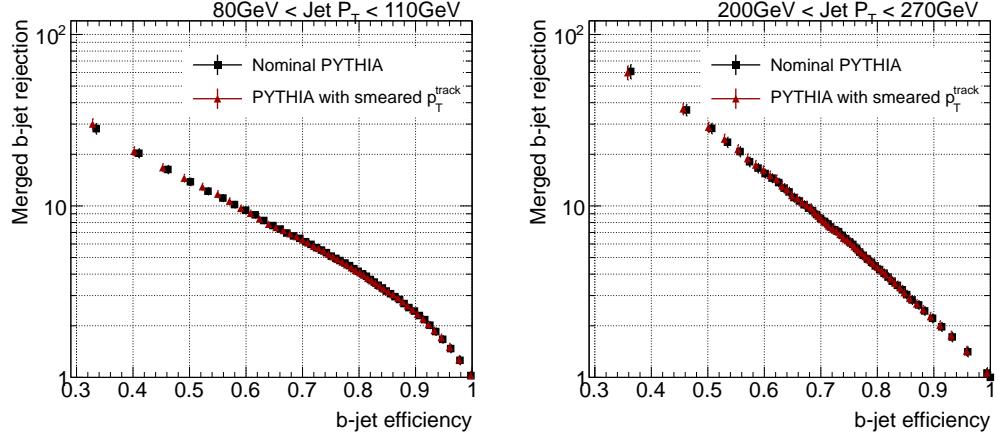


Figure 6.9: Rejection of merged b -jets as a function of the single b -jet efficiency showing the effect of the track momentum resolution uncertainty. It is found to be negligible with respect to the statistical uncertainty.

of the magnetic field. Its uncertainty propagates to the kinematic variables used in the double b -hadron jet tagger. In order to study this effect, track momenta are over-smeared according to the measured resolution uncertainties, before the track selection cuts are applied. The actual smearing is done in $1/p_T$, with an upper bound to the resolution uncertainty given by $\sigma(1/p_T) = 0.02/p_T$ [60]. The effect is found to be negligible, see Fig 6.9.

V. Jet energy scale and momentum resolution

The jet energy scale (JES) uncertainty for light jets reconstructed with the anti- k_t algorithm with distance parameter $R = 0.4$ and calibrated to the EM+JES scale is between $\sim 4\%$ at low p_T and $\sim 2.5\%$ for jets with $p_T > 60$ GeV in the central region [101]. In the case of b -jets, an additional uncertainty arising from the modelling of the b -quark production mechanism and the b -quark fragmentation was determined from systematic variations of the Monte Carlo simulation. The resulting fractional additional JES uncer-

tainty for b -jets has an upper bound of 2% for jets with $p_T \leq 100$ GeV and it is below 1% for higher p_T jets. To obtain the overall b -jet uncertainty this needs to be added in quadrature to the light JES uncertainty.

The systematic uncertainty originating from the jet energy scale is obtained by scaling the p_T of each jet in the simulation up and down by one standard deviation according to the uncertainty of the JES. The result is shown in Fig. 6.10a for a medium p_T bin. The effect on the likelihood performance is an average variation of 5% for the 50% and 60% efficiency working points.

The jet momentum resolution was measured for 2011 data and found to be in agreement with the predictions from the PYTHIA-based simulation [102]. The precision of this measurement, determined in p_T and η bins, is typically 10%. The systematic uncertainty due to the calorimeter jet p_T resolution was estimated by over-smearing the jet 4-momentum in the simulated data, without changing jet η or ϕ angles. The performance, shown in Fig. 6.10b, is found to globally decrease by 6%, without a particular p_T dependence.

The different contributions to the systematic uncertainty on the merged b -jet rejection are summarized in Table 6.2.

Although the likelihood training was performed in EM+JES calibrated jets, the performance of the tagger was also evaluated in jets calibrated with the LC+JES scheme, described in Section 4.1. A small degradation of the performance is observed, but comparable with the statistical uncertainties. A comparison of the performances is shown in Fig. 6.11 for two p_T bins, representative of the jet momentum range covered.

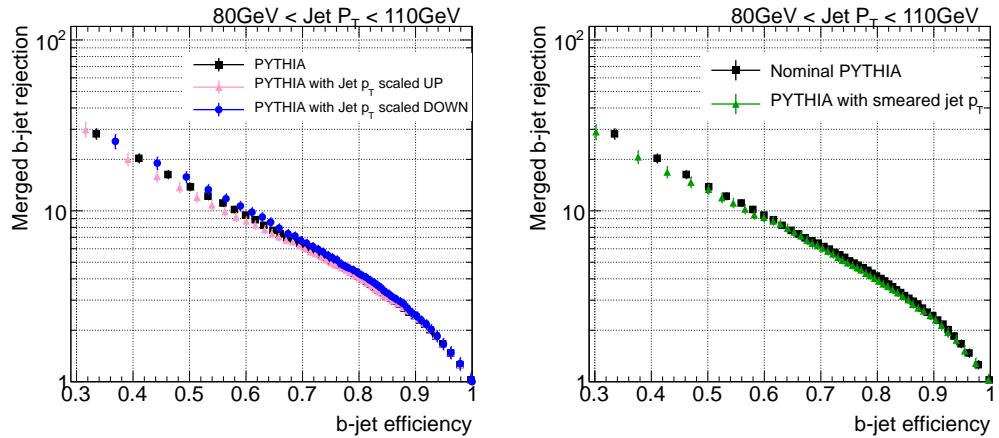


Figure 6.10: Rejection of merged b -jets as a function of single b -jet efficiency for (a) jets with smeared p_T and (b) for jets with varied energy scale compared to nominal.

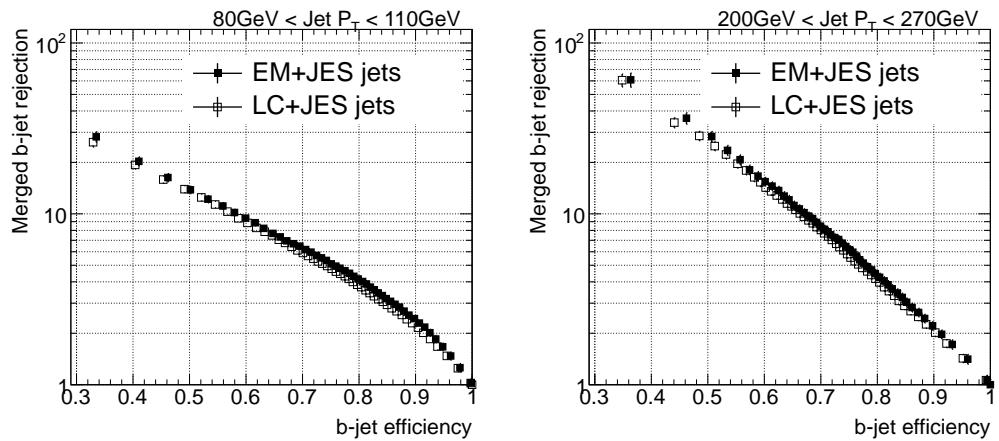


Figure 6.11: Rejection of merged b -jets as a function of single b -jet efficiency for jets calibrated to the EM+JES (LC+JES) scale, between 80 GeV and 110 GeV and 200 GeV and 270 GeV.

Systematic source	Uncertainty
pile-up	negligible
b -tagging efficiency	negligible
track reconstruction efficiency	4%
track p_T resolution	negligible
jet p_T resolution	6%
jet energy scale	5%

Table 6.2: Systematic uncertainties in the merged b -jet rejection (common to both the 50% and the 60% efficiency working points).

6.4 Other Monte Carlo generators

The development, training and performance determination of the tagger has been done using Monte Carlo events generated with the PYTHIA event simulator, interfaced to the GEANT4 based simulation of the ATLAS detector. An immediate question is what the performance would be if studied with a different simulation. In this section we investigate this question for the PYTHIA Perugia tune and the HERWIG++ event generators.

Fig. 6.12 shows a comparison of the likelihood rejection, at the 50% efficiency working point, between nominal PYTHIA and the alternative simulations as a function of the jet p_T . The larger errors are due to the reduced statistics available, which are even lower for the Perugia case than for HERWIG.

The performance in HERWIG shows a systematic trend, with agreement at low p_T and increasingly poor performances compared to PYTHIA as p_T grows. For the Perugia tune, on the other hand, there is no definite behavior,

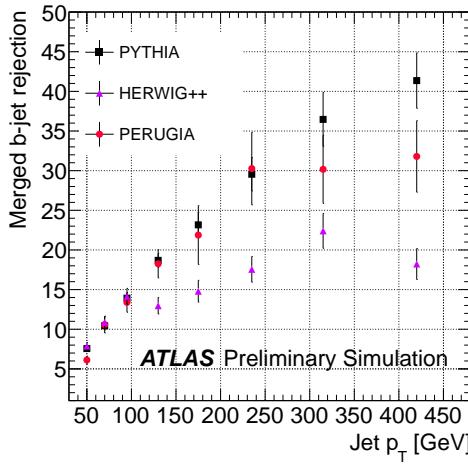


Figure 6.12: Rejection of merged b -jets as a function of jet p_T for different Monte Carlo generators, at the 50% efficiency working point.

with the performance fluctuating above or below the nominal simulation for different p_T bins consistently with the statistical uncertainties.

The reason for the systematic difference observed between the performances of PYTHIA and HERWIG can be traced to the extent with which jets are accurately modelled. Fig. 6.13 compares the measured jet track multiplicity distributions in b -tagged jets and the prediction from both simulations, for low and high p_T jets. It is observed that indeed HERWIG++ does not correctly reproduce the data, particularly at high p_T . The level of agreement is found to be better for track-jet width and the ΔR between the axes of the two k_t subjets in the jet, the two other variables used for discrimination.

6.5 Validation of the MVA output in data

In Section 5.3 the validation of the tracking variables in data from 2011 runs was presented, showing an excellent agreement between experimental data

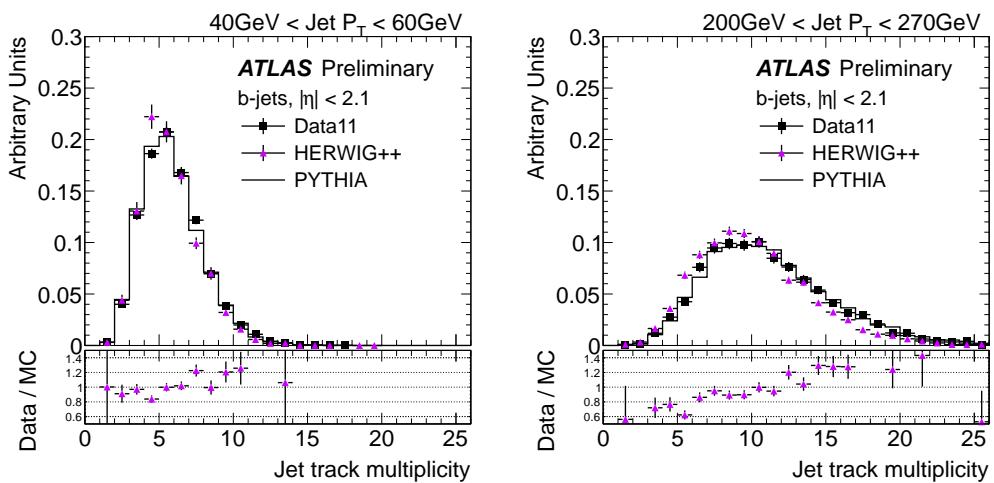


Figure 6.13: Distribution of the jet track multiplicity in 2 different jet p_T bins, for experimental data collected during 2011 (solid black points) and HERWIG++ events (solid violet triangles). The ratio data over HERWIG++ simulation is shown at the bottom of the plot. PYTHIA distribution is also shown for reference.

and the Monte Carlo simulation. In particular, the agreement obtained for the track multiplicity, the track-jet width and the ΔR between k_t axes - the three input variables - is shown in Fig. 6.14.

The likelihood presented in this chapter is an observable built out of the three selected variables and, for this reason, a similar level of agreement is expected for the output distributions in bins of the jet transverse momentum. Figures 6.14 shows the distribution of the likelihood output for different bins of jet p_T . The agreement is excellent.

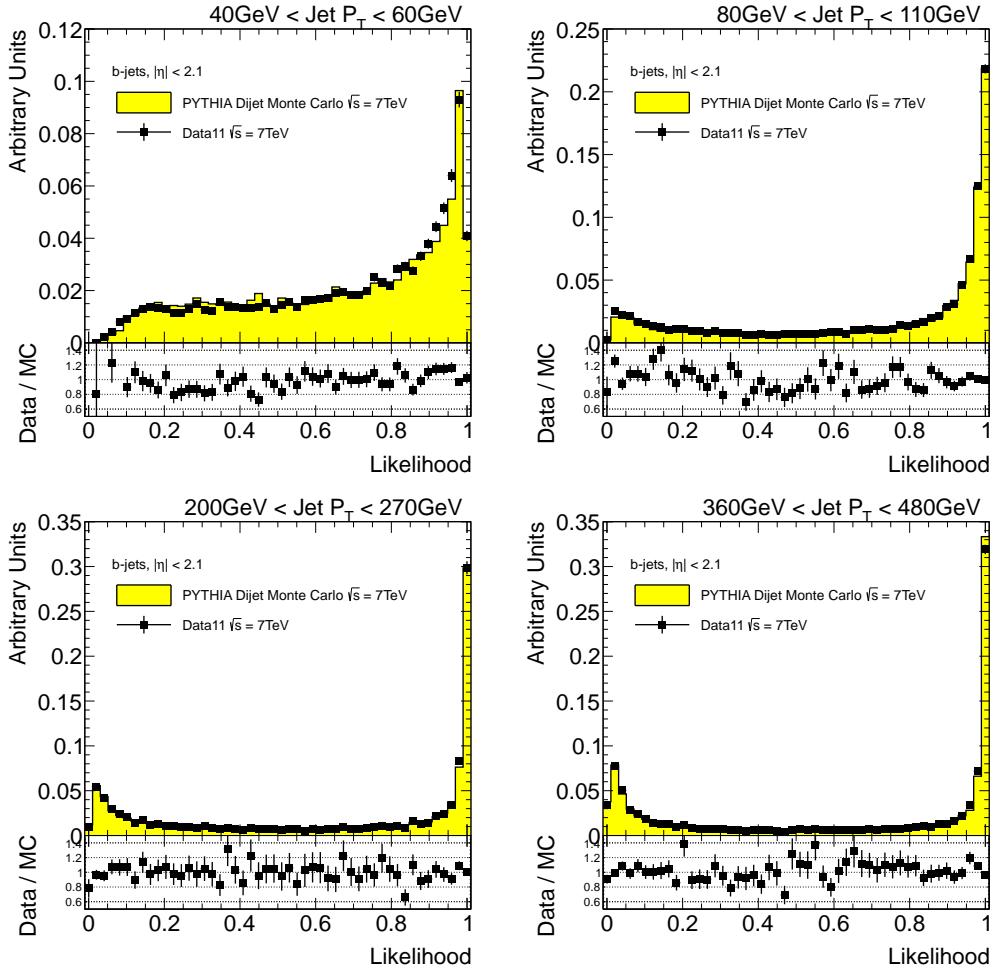


Figure 6.14: Distribution of likelihood output in different jet p_T bins, for experimental data collected by ATLAS during 2011 (solid black points), and simulated data (filled histograms). The ratio data over simulation is shown at the bottom of each plot.

Chapter 7

Summary and conclusions

In the course of the present thesis a new method allowing the identification of b -jets containing two b -hadrons was developed. The method exploits the expected kinematic differences between double b -hadron (“merged”) jets and single b -jets, combining a set of discriminating variables in a multivariate classifier. The differences between single and merged jets originate in the two-subjet structure of merged jets which, in QCD, are expected to arise mainly from a gluon splitting into a close-by $b\bar{b}$ -pair. Merged jets tend to have higher multiplicity and larger width. Several jet shape and substructure variables accounting for these envisaged characteristics were investigated in order to obtain the best single-merged discrimination. Due to the noisy environment of the hadron collisions at the LHC track-based variables were preferred over calorimeter variables.

The multivariate classifier was trained using simulated QCD events. A likelihood estimator was chosen among other multivariate methods for its robustness and simplicity. Based on discrimination power, correlation and pile-up dependence three input variables were selected for the tagger training: the jet track multiplicity, the track-jet width and the ΔR between the axes

of two k_t subjets in the jet. The performance of the tagger in Monte Carlo events was studied in bins of the calorimeter jet p_T , achieving a rejection of merged jets of over 95% (90%) for a 50% single b -jet efficiency for jets with $p_T > 150$ GeV ($p_T > 60$ GeV). Several sources of systematic uncertainties in the merged b -jet rejection were evaluated for the 50% and 60% single b -jet efficiency working points, being the most relevant the tracking efficiency and the jet energy scale and resolution with an average performance variation of 4%, 5% and 6%, respectively. Other contributions such as pile-up or the uncertainties in the track momentum resolution and the b -jet tagging efficiency resulted negligible.

The Monte Carlo distributions of the explored variables as well as the likelihood output were validated using experimental data corresponding to an integrated luminosity of 4.7 fb^{-1} recorded by the ATLAS experiment during 2011. The agreement between data and simulation is excellent.

This tool provides a handle to investigate QCD $b\bar{b}$ production and to reduce backgrounds in Standard Model physics analyses that rely on the presence of single b -jets in the final state, such as top quark physics (either in the $t\bar{t}$ or the single top channels) or associated Higgs production ($WH \rightarrow \ell\nu b\bar{b}$ and $ZH \rightarrow \nu\nu b\bar{b}$). These processes suffer from backgrounds that can be in part removed by a merged b -jet tagger. Jets containing a single b -quark or antiquark also enter in many BSM collider searches, the ability to distinguish single b -jets from jets containing two b -hadrons is thus here of wide application to reduce SM backgrounds giving rise to close-by $b\bar{b}$ pairs.

In order to expand up the results presented here, and to make further advancements in the implementation of the tagger in physics analyses the following improvements should be made: the extension to non-isolated jets using the concept of ghost-particle matching and active area of a jet for track-

to-jet association and labeling and the calibration of the tagger with data. Nontheless, the study presented in this thesis demonstrates that jet substructure variables can provide a good handle for gluon splitting identification in physics searches within ATLAS.

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