

UNIVERSITY OF OKLAHOMA

GRADUATE COLLEGE

SEARCHING FOR SUPERSYMMETRIC PARTICLES AT THE LARGE HADRON
COLLIDER USING THE ATLAS DETECTOR

A DISSERTATION

SUBMITTED TO THE GRADUATE FACULTY

in partial fulfillment of the requirements for the

Degree of

DOCTOR OF PHILOSOPHY

By

OTHMANE RIFKI
Norman, Oklahoma
2016

SEARCHING FOR SUPERSYMMETRIC PARTICLES AT THE LARGE HADRON
COLLIDER USING THE ATLAS DETECTOR

A DISSERTATION APPROVED FOR THE
HOMER L. DODGE DEPARTMENT OF PHYSICS AND ASTRONOMY

BY

Dr. Brad Abbott, Chair

Dr. Peter Barker, Outside Member

Dr. Mike Strauss

Dr. Chung Kao

Dr. Eric Abraham

To ...

Table des matières

Acknowledgments	v
Abstract	vi
Preface	vii
1 Introduction	1
2 Theoretical Background	11
2.1 The Standard Model of Particle Physics	11
2.2 Supersymmetry	11
2.3 Phenomenology at the LHC	11
3 Experimental Apparatus	12
3.1 The Large Hadron Collider	12
3.2 A Toroidal LHC Apparatus (ATLAS)	16
3.2.1 Co-ordinate System	17
3.2.2 Inner detector	18
3.2.3 Calorimeters	18
3.2.4 Muon System	18
3.3 ATLAS Detector	18
3.4 ATLAS Trigger and Data Acquisition System	19
3.4.1 Hardware Trigger (L1)	19
3.4.2 Dataflow Challenges in Run-2	20
3.4.3 ATLAS Dataflow Design	21
3.5 Event Simulation	24
3.6 Reconstruction and Identification Techniques	26
3.6.1 Electrons	27
3.6.2 Muons	28
3.6.3 Jets	29
3.6.4 Heavy flavor	30
3.6.5 Missing transverse momentum	31
4 The Region of Interest Builder	43
4.1 Overview	43
4.2 VMEbus based RoIB	44
4.2.1 Hardware implementation	44
4.2.2 System Performance and Evolution	46
4.3 PC based RoIB	47
4.3.1 The Common Readout Receiver Card	48
4.3.2 Readout System Firmware and Software	49
4.3.3 RoIB Software	50
4.4 Prototype Tests	51
4.4.1 Standalone Tests	52
4.4.2 Full System Tests	52
4.5 Online Performance in Run-2	55

5 Analysis Strategy	57
5.1 Overview	57
5.2 Phenomenology (Benchmarking Models)	59
5.3 Dataset and Simulated Event Samples	67
5.3.1 Collision Data	67
5.3.2 Simulated Event Samples	68
5.4 Event Selection	75
5.4.1 Pre-selection and event cleaning	75
5.4.2 Trigger strategy	77
5.4.3 Object definition	78
5.4.4 Data-MC comparisons	82
5.5 Signal Regions	83
5.6 Analysis Acceptance and Efficiency	88
6 Data-driven Background Estimation Techniques	100
6.1 The problem of fakes	100
6.2 Common processes for faking leptons	101
6.3 The Monte Carlo Template Method	103
6.3.1 Motivation	103
6.3.2 Description of the method	104
6.3.3 Correction factors	105
6.3.4 Control regions	106
6.4 Matrix Method	110
6.4.1 Events with one object	111
6.4.2 Dynamic matrix method	113
6.4.3 Propagation of uncertainties	118
7 Background Estimation	119
7.1 Overview	119
7.2 Irreducible Backgrounds	119
7.2.1 Expected yields in the signal regions	119
7.2.2 Validation regions	120
7.2.3 Validation of irreducible background estimates	126
7.3 Reducible Backgrounds	126
7.3.1 Charge-flip Background	126
7.3.2 Matrix Method	135
7.3.3 MC Template Method	148
7.3.4 Reducible Background Validation	150
7.4 Systematic uncertainties	154
7.4.1 Theoretical Uncertainties	154
7.4.2 Experimental Uncertainties	157
8 Statistical Treatment	170
8.1 Likelihood Function	171
8.2 Limit Setting Procedure	173
8.2.1 Profile Likelihood Ratio	173
8.2.2 The p -value	174
8.2.3 The CL_s Prescription	176

8.2.4	Approximate Sampling Distributions	178
8.3	Statistical Implementation	179
9	Results and Interpretation	181
9.1	Predictions and Observations	181
9.2	Statistical Interpretation	184
9.2.1	Model independent discovery and upper limits	187
9.2.2	Model dependent exclusion limits	187
10	Conclusions	196
A	Auxiliary material	197
A.1	Signal region with best exclusion	197
A.2	Upper limit on cross section	199
A.3	Signal region cutflow	201
A.4	Acceptance and Efficiency	201
B	Validation of background estimates	219
C	HistFitter Tables	220
	Bibliographie	221

Acknowledgments

I would like to acknowledge...

Abstract

My abstract ...

Preface

My preface here...

Over the past five years, I have always been asked by family, friends, and people I meet, what do you do ? I have written the introduction with a general audience in mind to answer this question.

Chapitre 1

Introduction

Historical Background

Of what is the universe made? A question that has intrigued the human curiosity since the dawn of time. Today, we are confident that we do not know the answer to this question. However, a lot of progress has been made with the aim of reducing the diversity of the physical phenomena observed around us to a limited number of constituents following fundamental principles. Over two thousands years ago, the ancient Greeks postulated that all is made of Earth, Air, Fire and Water. Fast-forward to the end of the 19th century, Mendeleev and others made the astonishing remark that by organizing the relative atomic masses of chemical elements, elements with similar chemical properties followed a pattern. The periodic table of elements was born. The predictive power of the periodic table led to the anticipation of new elements that were later discovered. However, the table lacked compactness and necessitated a more fundamental underlying structure that could connect the different elements together.

At the turn of the 20th century, several important discoveries established the existence of the atom and its constituents. The atom was formed by electrons bound via the electromagnetic force to a nucleus, where all the mass resides. The nucleus itself is formed from protons and neutrons that are “glued” together by the strong nuclear force (or strong force). These elements formed the underlying substructure that explained qualitatively the systematic organization of the

periodic table. Moreover, quantum ideas were applied to the atom offering a quantitative description of the origin of structure in atoms and molecules, including the chemical elements and their properties. The decades that followed refined our understanding of the composition of matter through a series of experimental results. By studying the collisions of protons and neutrons in the 50's and 60's, a plethora of new particles were discovered which belonged to the same family as the proton and neutron, called *hadrons*. These particles could not be all elementary¹. By invoking the same argument that atoms were composite based on Mendeleev's table, a new layer of structure was unfolded to reveal the existence of *quarks* as basic constituents of all hadrons. Six types of quarks were discovered over the years with the latest, the top quark, being the most massive elementary particle discovered at Fermilab in Chicago, Illinois in 1995 [1, 2].

The observation of the continuous energy spectra in the radioactive β decay of nuclei led to the discovery of neutrinos to remedy the energy conservation law in the decay. Neutrinos are very light neutral particles that interact via the weak nuclear force responsible for radioactivity and nuclear fusion, the process that powers the stars. Electrons and neutrinos had other relatives collectively called *leptons*. The quarks and leptons are referred to as *fermions* and have a half integer spin, an intrinsic property of elementary particles. The strong, weak, and electromagnetic interactions are mediated by gluons, W and Z bosons, and photons, respectively. These particles have an integer spin and are called *bosons*. The latest addition to the known elementary particles happened in 2012 with the discovery of a new

1. Elementary particles refer to particles that cannot be decomposed into further constituents.

boson, the Higgs boson, that allows the quarks and leptons and the W and Z bosons to acquire mass[3, 4].

The Standard Model

The physics of elementary particles became the most ambitious and organized attempt to answer the question of what the universe is made out of. Through a mixture of both theoretical insight and experimental input, we now know that everything we see in our daily life is formed from quarks and leptons that interact via the strong, weak, and electromagnetic forces.² The forms of these forces are determined from basic principles of symmetry and invariance. As a result, a theoretical framework was constructed to synthesize all these developments in a quantitative calculational tool that became known as the *standard model of particle physics* (SM). The only inputs needed by the SM are the interaction strengths of the forces and quark and lepton masses to make very accurate predictions about the behavior of elementary particles. Over the past 30 years, the SM has been vigorously tested by many experiments and has been shown to accurately describe particle interactions at the highest energies produced in the laboratory. Yet, we know it is not the complete story.

2. The fourth fundamental force of gravity is extremely weak and only acts at the macroscopic scale.

Limitations

In 1933, an observation of the Coma Cluster by Zwicky suggested that the galaxies in the cluster were moving too fast to be explained by the luminous matter present[5]. The same observation was repeated when looking at the rotation speeds of individual galaxies which suggested a dark component of mass, dark matter. Several independent measurements established that not only dark matter is not made out of baryons but it is more abundant. For example, anisotropies in the cosmic microwave background, a radiation left over from the Big Bang, were consistent with quantum fluctuations from an inflationary epoch [6, 7]. These fluctuations encoded details about the density of matter in the form of cosmological parameters as they traveled through space and time to reach our experiments. The astonishing conclusion was that the universe has nearly five times as much dark matter as ordinary matter [8]. The supernovae surveys gave direct evidence for an accelerating universe [9], a view that was cemented by the measurement of cosmological parameters [10, 11] which led to the startling discovery that most of the energy density of the universe is in the form of an unknown negative-pressure, called dark energy [12]. There is an extensive program of experiments which will probe the dark energy. Astrophysics and cosmology told us about the existence of dark matter and measured its density to a remarkable precision. Particle physics holds the hope to uncover what dark matter is. In short, all experimental evidence are consistent with a universe constructed of

- baryons (everyday matter) : $\sim 5\%$

- dark matter : $\sim 20\%$
- dark energy : $\sim 75\%$
- neutrinos, photons : a tiny fraction

Today, we are in front of many puzzles related to our view of the universe.

Everything we know of, namely all particles of the standard model, constitutes only 5% of the energy budget of the universe. The universe is also predominately composed of matter as opposed to anti-matter even though at the start of the universe, they were in equal amounts. The standard model describes the content of everyday matter and how it interacts but without telling us why it is that way. Moreover, the standard model only describes these phenomena up to an energy scale of $\mathcal{O}(100)$ GeV, weak scale. Beyond this scale lies the realm of phenomena not described by the standard model that extend all the way to the Planck scale of $\mathcal{O}(10^{19})$ GeV. There is no mechanism to generate mass for neutrinos in the standard model. Last but not least, the standard model does not incorporate gravity, the fourth fundamental force. The SM is unable to account for these observed features in the universe. Thus, there is a need for a theory beyond the standard model.

Supersymmetry

One of the most prominent extensions of the standard model, that addresses many of the shortcomings mentioned above, is a theory based on a new symmetry, called supersymmetry. This symmetry is between the matter particles, fermions, and particles whose exchange mediates the forces, bosons. Our current description

of the world treats fermions and bosons differently. Supersymmetry puts forward the idea that fermions and bosons can be treated in a fully symmetric way. In other words, if we exchange fermions and bosons in the equations of the theory, the equations will still look the same. An immediate consequence of the theory is that every standard model particle will have a “superpartner”. As a result, we can design experiments to search for these supersymmetric particles. The work presented in this dissertation is about the search for supersymmetric particles with a specific signature. The many benefits of supersymmetry will be discussed later but here it is worth mentioning two important features of the theory : it unifies the three interactions, electromagnetic and strong and weak forces, at very high energies and it provides a dark matter candidate particle. Now that we understand what we are trying to do, it is time to address the question of how to do it.

Experimental techniques

The human eye can resolve pieces of dust up to 10^{-5} m. The subatomic distances we are interested in probing range from 10^{-15} m, the size of a proton, down to 10^{-18} m, the size of a quark. Instruments are needed to expand our senses to probe these very small scales. For instance, light microscopes can reveal what things are like down to 10^{-6} m, the scale of bacteria and molecules. A special type of microscope is needed to probe smaller distances, a particle accelerator. The basic idea is that in order to see an object, a wave must scatter off this object and must have a wavelength smaller than the object being probed. Since particles have

a wavelike character, they can be used to probe ever shorter distances according to

$$E = \frac{hc}{\lambda}$$

where E is the energy of the particle, λ is its wavelength, and $hc \sim 10^{-6} eVm$.

As a result, the higher the speed of the particles, the greater their energy and momentum and the shorter their associated wavelength. Modern accelerators can generate energy in the TeV scale and thus probe a distance of 10^{-18} m. All the development that we have made describes phenomena happening at distances larger than about 10^{-18} m. Thus, it is possible that electrons and quarks have some structure which are beyond our ability to resolve in experiment. For this reason, we consider them as not having any deeper structure, pointlike objects.

Over the last century, beams of particles were used to study the composition of matter. Initially, beams originated from phenomena that were already occurring, such as alpha and beta particles, coming from radioactive decays, and cosmic rays. Cosmic rays are much more energetic than what we can produce in the laboratory today, however, they came at random and with a low intensity. Instead, high energy particle accelerators were used to deliver high intensity beams of electrons, protons, and other particles under controlled conditions. For this reason, particle physics is also known as high energy physics. By colliding two sufficiently energetic particles, new particles will be created according to Einstein's equation $E = mc^2$ (or more generally $E = \sqrt{(mc^2)^2 + (pc)^2}$), where energy can be exchanged for mass, and vice versa, the exchange rate being c^2 , the square of the speed of light. For example, an electron has a mass of 0.5 MeV and can only be created in a

pair of particle and its antiparticle counterpart, thus 1 MeV of energy is needed for an electron–positron pair to be produced at rest. Energies in the TeV range were present about a billionth of a second after the Big Bang. In other words, by colliding high energy particles, it is possible to recreate momentarily conditions similar to those of the universe when it was newly born. At such energies, particles and antiparticles were created, including exotic forms no longer common today. Most of the particles generated in these collisions are extremely short lived with lifetimes less than 10^{-20} seconds, producing radiation and decaying to stable particles, such as electrons and quarks, that make up most of what we see today. One of the exotic forms of matter that may exist is supersymmetry. The search for evidence for supersymmetric particles using data collected at a high energy particle accelerator is the subject of this dissertation.

The Large Hadron Collider (LHC) is the world's most energetic particle accelerator and the pinnacle of colliding beam technology. Is it located at CERN, the European Laboratory for Particle Physics³, near Geneva, Switzerland. The LHC accelerates counter rotating beams of protons (and even atomic nuclei) to 99.9999991% the speed of light in a 27 km ring reaching an energy of 6.5 TeV per beam. The most important part of the design is the magnets, cooled by the largest cryogenic system in the world to 1.9 K (-271.3 °C), that keep

3. The acronym comes from French “Conseil Européen pour la Recherche Nucléaire” which was established to do fundamental physics research. In 1952, this research concentrated on understanding the atom and its nucleus, hence the word “nuclear”. Today, our knowledge goes deeper than the nucleus which motivates the modified name.

the protons on track and bring the counter-rotating needle-like beams to meet head on 40 million times per second. The debris of each collision fly off in all directions, briefly producing less common exotic forms of matter captured by large particle detectors in the form of “snapshots” of these collisions, called events. The teams of scientists analyze these events to identify the different particles that were produced and reconstruct the full collision process. With this information, it is possible to make precision measurements of rare standard model processes, like the production of the Higgs boson, or search for beyond the standard model physics, like evidence for supersymmetry. ATLAS is one of the general-purpose particle detectors at the LHC that supplied the events analyzed in this dissertation to search for supersymmetry. The ATLAS detector is the largest-volume particle detector ever built – the size of a seven-story building 46 meters high and 26 meters in diameter, weights 7000 tonnes, and able to measure particle trajectories down to 0.01 meters. Bunches of protons would pass through each other at the heart of the ATLAS detector 40 million times per second. Each time they cross there are on average 25 proton-proton collisions, leading to about a billion proton collisions per second. The data generated in these collisions amounts to about 60 terabytes per second, an amount far beyond what is technologically possible to store. In fact, the processes of interest are extremely rare. For example, the Higgs boson is produced once in 20 million million collisions. In more practical terms, a Higgs boson might appear once a day during the LHC operations. ATLAS has a big computational challenge to recognize this one Higgs event and record it to tape out of 35 million million other collisions each day. The topic of this

dissertation is to search for supersymmetric particles that are even rarer and thus more challenging to look for.

This dissertation will give detailed explanation on how we searched for supersymmetric particles using the ATLAS detector. First, the motivation behind the work will begin with an overview of the standard model of particle physics and supersymmetry in Chapter 2 followed by the design of the ATLAS detector at the LHC in Chapter 3. The Region of Interest Builder that processes every event recorded by ATLAS is covered in Chapter 4. The detailed description of the search starts in Chapter 5 covering the basic analysis strategy and the supersymmetric models considered. The most important part of the analysis is the estimation of standard model and detector backgrounds with novel techniques developed by the author and covered in Chapters 6 and 7. The statistical methodology and interpretation of the results is presented in Chapters 8 and 9. This analysis represents an important search for supersymmetric particles with the early data-set collected by ATLAS at a new center of mass energy of 13TeV. The strength of the search lies in exploring regions of the parameter space with a small mass difference between the supersymmetric particles, regions that are difficult to probe with other searches for new physics.

Chapitre 2

Theoretical Background

2.1 The Standard Model of Particle Physics

2.2 Supersymmetry

2.3 Phenomenology at the LHC

Chapitre 3

Experimental Apparatus

3.1 The Large Hadron Collider

The Large Hadron Collider (LHC) is the largest particle accelerator and collider in the world. The LHC is built in a circular tunnel 27 km in circumference that is buried between 50 m to 175 m underground and straddles the Swiss and the French borders at CERN. The LHC is a synchrotron that accelerates two counter-rotating beams of protons to 6.5 TeV then brings them into head-on collisions at the center of four large detectors : ALICE, ATLAS, CMS, and LHCb. The center of mass energy of the proton-proton (or pp) collision is $\sqrt{s} = 13$ TeV, the energy of the collision data analyzed in this dissertation. The beam itself has a total energy of 336 MJ requiring an accurate and careful steering of the beam at all times. This is achieved by a strong magnetic field generated by superconducting magnets that guide the protons around the accelerator. There are 1232 dipoles magnets, each 15 meters long operating at 1.9 K and generating a magnetic field of 8.33 T. The dipoles comprise of 7600 km superconducting cable which is formed from filaments of Niobium-titanium (NbTi).

A complex of smaller accelerators boost the protons before injecting them to the LHC, the last accelerator in the chain as shown in Figure 3.1. Protons, obtained from hydrogen atoms, start their in a linear accelerator called Linac2. The Linac2 accelerates the protons to 50 MeV. Then, they are injected into the

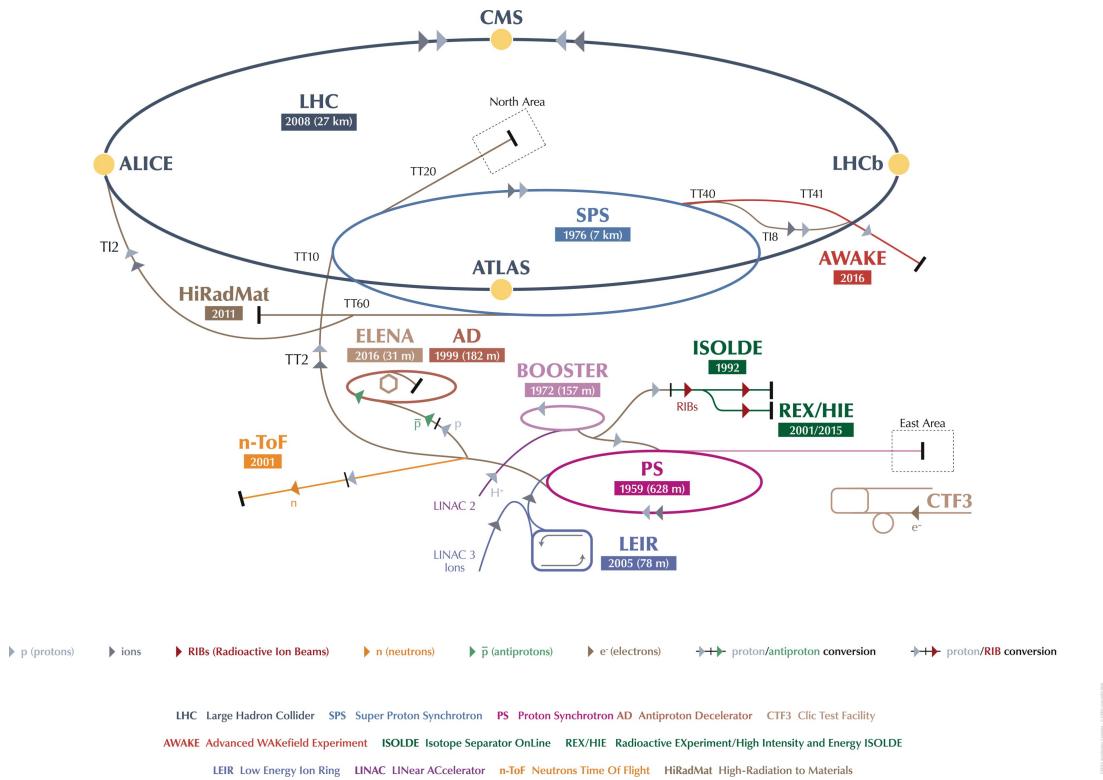


FIGURE 3.1: The CERN accelerator complex composed of a chain of particle accelerators with the LHC as the last ring (dark blue line) [?].

PS Booster, which accelerates them to 1.4 GeV. After the PS Booster, the protons are sent to the Proton Synchrotron where they are accelerated to 25 GeV. They are then sent to the Super Proton Synchrotron (SPS) where they are accelerated to 450 GeV. At this stage, the protons are injected into the LHC and accelerated to the target energy of 6.5 TeV per proton. The beams are then focused at each of the interaction points to produce proton-proton collisions. Under normal operating conditions, the colliding beams will circulate for $\mathcal{O}(10^{19})$ hours at a time.

The protons are grouped in “bunches” when circulated in the LHC as a result of the acceleration scheme. In normal operation of the LHC, each proton beam

has 2808 bunches, with each bunch containing about 100 billion protons. These bunches are a few centimetres long and a few millimeters wide when they are far from a collision point but squeezed to about 16 micrometres when they collide. The rate of their interaction is defined in terms of the luminosity, a measure of the number of collisions produced by the accelerator. Generally, the event rate $\frac{dN}{dt}$ of a physics process with cross section σ is

$$\frac{dN}{dt} = \sigma \mathcal{L} \quad (3.1)$$

where the constant of proportionality, \mathcal{L} , is called the instantaneous luminosity, and has units of $cm^{-2}s^{-1}$. The LHC has exceeded its design luminosity of $10^{34}cm^{-2}s^{-1}$ or $10nb^{-1}s^{-1}$ ($1 \text{ barn} = 10^{-24}cm^2$) by almost 40% as shown in Figure 3.2a. Given the total inelastic cross section of 60 mb, the collision rate of protons is then $\sigma\mathcal{L} \sim 10^9$ Hz : a billion proton interactions per second. The integral of the instantaneous luminosity, $L = \int \mathcal{L} dt$, refers to the amount of data collected. The large integrated luminosities allow for the study of rare processes, such as the search for supersymmetric particles. Figure 3.2b shows the data sets collected by the LHC where the data collected in 2015 and 2016 at $\sqrt{s} = 13$ TeV is the basis of the work presented in this dissertation.

The other important characteristic of the LHC is that multiple pp interactions occur at every bunch crossing. This quantity is correlated with the instantaneous luminosity as can be seen by comparing Figure ?? and Figure ?? . Figure 3.3b shows that the mean number of interactions per bunch crossing was 25 in 2016 with peak number of interactions reaching up to 50. This causes a computational challenge

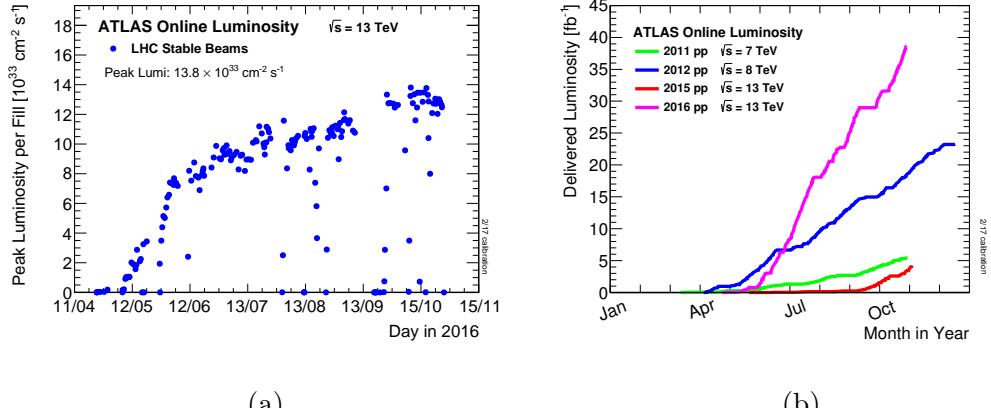


FIGURE 3.2: (a) The peak instantaneous luminosity delivered to ATLAS in 2016 and (b) the cumulative luminosity delivered to ATLAS between 2011 and 2016, during stable beams for pp collisions

in reconstructing the physics objects coming from one interesting interaction.

A typical hard scattering of two protons has a large impact parameter leading

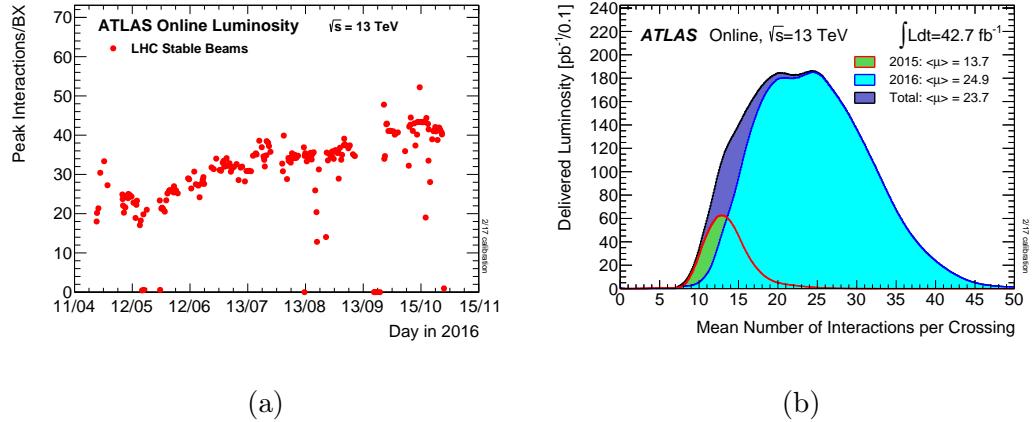


FIGURE 3.3: (a) the peak number of inelastic collisions per beam crossing during 2016 and (b) the mean number of these collisions per crossing for 2015 and 2016, during stable beams for pp collisions

to low momentum particles in the final state. This type of collisions are known

as “minimum bias” collisions and are considered as background to the more spectacular hard scattering that is typical of an interesting event. The minimum bias is generally not well understood since it comes from nonperturbative QCD and is estimated using the LHC data.

3.2 A Toroidal LHC Apparatus (ATLAS)

ATLAS (**A** Toroidal **L**H**C** Apparatu**S**) is a multi-purpose particle detector located at one of the LHC interaction points 100 meters underground. It is the largest particle detector ever built with a weight of about 7000 tonnes, a length of 44 m, and a diameter of 25 m as shown in Figure ???. It is designed to probe Higgs physics, QCD, flavour physics, as well as a multitude of beyond Standard Model (BSM) physics scenarios including supersymmetry.

ATLAS covers a solid angle of almost 4π to capture as much information from the collisions as possible. It is composed of multiple layers of detectors to ensure that all particles produced in the collision are identified and measured with high accuracy. These subsystems are shown in Figure ???. The first detector resides in the part closest to the LHC pipe and is composed of silicon tracking sensors designed to reconstruct the paths of charged particles. Next, comes the electromagnetic and hadronic calorimeter cells that measure the energy of particles. Last comes the muon spectrometer in the outermost part of the detector to detect muons since they penetrate the calorimeters. An axial magnetic field of 2 T is applied across the inner detector while a toroidal magnetic field of approximately

0.5 T is applied across the muon detectors. The remainder of this chapter will describe in more details these detectors.

3.2.1 Co-ordinate System

The common coordinate system of ATLAS is right-handed Cartesian, with its origin at the nominal interaction point. The axes are oriented such that the x -axis is pointing towards the centre of the LHC ring, the y -axis is directed vertically upward, and the z -axis defines one of the beam directions. The (x, y) plane defines the transverse plane, usually represented by polar coordinates (r, ϕ) with $\phi = 0$ on the x -axis. The polar angle θ is replaced by the pseudorapidity

$$\eta = -\ln \left(\tan \left(\frac{\theta}{2} \right) \right). \quad (3.2)$$

It is named after the rapidity since it yields the same quantity for massless particles

$$y = \frac{1}{2} \ln \left(\frac{E + p_Z}{E - p_Z} \right), \quad (3.3)$$

which is invariant under boosts in the z -direction. It is common to describe the separation between two physical objects in the detector by

$$\Delta R = \sqrt{\Delta\eta^2 + \Delta\phi^2} \quad (3.4)$$

3.2.2 Inner detector

3.2.3 Calorimeters

LAr Calorimeters

Tile Calorimeters

3.2.4 Muon System

3.3 ATLAS Detector

TABLE 3.1: Luminosity weighted relative fraction of good quality data delivery efficiencies (%) by the various components of the ATLAS detector and trigger subsystems during LHC stable beams in pp collisions at $\sqrt{s} = 13$ TeV with 25 ns bunch spacing between April-October 2016, corresponding to a recorded integrated luminosity of 35.9 fb^{-1} . The toroid magnet was off for some runs, leading to a loss of 0.7 fb^{-1} .

Inner Tracker			Calorimeters		Muon Spectrometer					Magnets		Trigger
Pixels	SCT	TRT	LAr	Tile	MDT	RPC	CSC	TGC	Solenoid	Toroid	L1	
98.9	99.9	99.7	99.3	98.9	99.8	99.8	99.9	99.9	99.1	97.2	98.3	
Good for physics : 93-95% (33.3-33.9 fb^{-1})												

3.4 ATLAS Trigger and Data Acquisition System

The ATLAS detector's data acquisition system, illustrated in Figure 3.9, makes use of a multi-tiered trigger to reduce bandwidth from the LHC proton bunch crossing rate of 40 MHz to the 1 kHz written to disk [13, 14]. The first tier (Level-1 or L1) [15], implemented in real time with custom electronics, makes an early event selection to determine if any objects of interest are present and reduces the data flow to 100 kHz. The second tier, referred to as the High Level Trigger (HLT) [16], is implemented on a commodity computing cluster running custom triggering software. The HLT uses information from the hardware based L1 system to guide the retrieval of information from the Readout System (ROS) [17].

3.4.1 Hardware Trigger (L1)

The L1 trigger has access to raw data from the calorimeters and the muon system. The L1 calorimeter trigger (L1Calo) uses reduced-granularity information from 7200 trigger towers of the calorimeters. These trigger towers are divided in a $\Delta\eta \times \Delta\phi$ space by 0.1×0.1 over most of the calorimeter, and larger in the forward region. A decision is made based on the multiplicities and E_T thresholds of the objects identified by the L1Calo algorithms : EM clusters, τ -leptons, jets, missing transverse energy, scalar sum E_T , and total transverse energy of the L1 jets. The L1 muon trigger (L1muon) uses measurements of the trajectories of muons in the RPC and TGC trigger chambers, located in the barrel and end-cap regions of the muon spectrometer. The multiplicity of the various muon p_T thresholds is input

to the trigger decision.

The central trigger processer (CTP) combines results from the L1Muon and L1Calo triggers to issue an overall L1 accept or reject decision. To facilitate this task, the CTP programs up to 256 configurations that consist of various combinations of E_T and p_T requirements, or thresholds. The CTP has the capability of implementing different isolation criteria to the different objects such as the L1 EM clusters. A trigger menu is implemented as a collection of L1 items, each containing a logical combination of one or more configured L1 thresholds. For example, the item L1_EM30i refers to an event requiring at least one isolated EM object with $E_T > 30\text{GeV}$. If the rates of a particular object is high , such as EM objects with low momentum, a prescale factor α is applied to the L1 item in the menu, where only 1 in α events is passed to the HLT. The L1 prescales are generally adjusted to maintain the optimal use of the allocated bandwidth for an L1 item during data-taking since the luminosity drops over the course of a run.

The L1 trigger has a $2.5\ \mu\text{s}$ latency where the data fragments are hold in pipeline buffers located within detector-specific front-end electronics. Once the CTP issues an accept, the data is pushed to detector-specific Readout Drivers (RODs), then transferred to the Readout System (ROS). The rest of the chain is described next.

3.4.2 Dataflow Challenges in Run-2

The function of the DAQ system is to efficiently buffer, transport, and record the events that were selected by the trigger system. Its performance is affected

by the instantaneous luminosity that leads to busy events with multiple proton-proton interactions occurring in each bunch crossing, referred to as pileup. The high pileup results in a higher data volume collected by the detector that needs to be processed at the required rate to avoid exerting back-pressure on the L1 system. In Run 2, the LHC has exceeded the designed instantaneous luminosity of $10^{34} \text{ cm}^{-2} \text{ s}^{-1}$ leading to pileup of $<\mu> = 30$ or more as shown in Figure 3.11. The L1 accept rate has also increased from 75 kHz in Run 1 to 100 kHz in Run 2 and the average output rate of the data logger system has increased from 400-600 Hz in Run 1 to about 3 kHz with 1.5 kHz for physics data. Moreover, there were new detectors that were added in Run 2 (Insertable B-layer (IBL), L1 topological trigger, Fast Tracker (FTK))[19] leading to an increase of 20% in the number of readout channels. To cope with these changes, the ATLAS TDAQ system was upgraded during Run-2 simplifying its architecture and increasing its flexibility. To be able to deliver more rate to the High Level Trigger (HLT), the upgrade also targeted the Readout System (ROS)[20]. For the same reason the two level of the HLT system were collapsed into a single level which made the system more flexible allowing for incremental data retrieval and analysis. The dataflow network system was re-designed to increase its capacity and simplify its architecture[21].

3.4.3 ATLAS Dataflow Design

In Run 1, the farm was subdivided to several slices, with each slice managed by a dedicated supervisor. This layout has been dropped in favor of global management by a single farm master operating at 100 kHz referred to as the HLT

supervisor (HLTSV). The Region of Interest Builder (RoIB) that assembles the RoIs previously implemented on a VMEbus system is now integrated with the HLTSV and the ROI building done in software. Chapter 4 is dedicated to the work of the author in the RoIB evolution. The change in the HLT architecture from two to one level required re-writing the HLT software and algorithms in such a way that each node in the farm can perform all processing steps. The handling of these processing steps is done by a single Data Collection Manager (DCM) process running on each HLT node to manage the L1 RoIs, the dataflow between the ROS and the HLT processing units (HLTPU), the event building processes, and the data logging. In the new architecture, the computing resources are managed more efficiently by balancing the utilization of all cluster nodes depending on the active HLT algorithms and by sharing the HLT code and services to reduce memory and resource usage.

The dataflow network shown in Figure 3.12 was simplified and upgraded to handle a larger data volume. A single network is used for ROI based access from the ROS, event building in the HLT processing nodes, and sending data for logging. A 10 GbE connectivity has been adopted throughout the dataflow system resulting in a factor of four increase in bandwidth between the data loggers and the permanent storage, and a 4×10 GbE output from each ROS PC to the core routers. The HLTSV and the HLT racks are all connected directly to each of the two core routers via 2×10 GbE connection. Each HLT rack is hosting up to 40 nodes connected by 2×1 GbE to the top-rack switches. The capacity of the routers can accommodate an increase in the number of HLT server racks and

ROS PCs by a factor of two, which will be needed when the system scales as run conditions change. The core routers also provide load balancing and traffic shaping protocols [21] to distribute the data throughout the system more evenly. A duplication of core routers provide link redundancy at every level in case of link or switch failures.

To take advantage of multi-core architectures, the dataflow software is using multi-threaded software design for CPU consuming operations. The Input/Output of the dataflow is based on asynchronous communication using industry standard libraries such as the Boost : :ASIO library. All the ATLAS software suite was switched to exclusively 64 bit operation in 2016.

In summary, the elements of the Run-2 ATLAS dataflow are :

- The Readout System (ROS) buffers front-end data from the detectors and provides a standard interface to the DAQ system.
- The Region of Interest Builder (RoIB) receives teh L1 trigger information from the RoIs and combines the information for the HLT supervsior.
- The HLT Supervisor (HLTSV) can handle the input from the RoIB and manage the HLT farm of about 2000 machines at over 100 kHz.
- The Data Collection Manager (DCM) handles all Input/Output on the HLT nodes, including ROI requests from the HLT and full event building.
- The HLT processing units (HLTPU) run the actualy HLT algorithms which are forked from a single mother process to maximize memory sharing.
- The Data loggers or SubFarm Output (SFO) are responsible for savind the accepted events to disk, and sending the files to CERN permanent storage

infrastructure.

3.5 Event Simulation

In order to interpret the LHC data, it is essential to compare the observations to the expected outcomes from a physical model, typically the standard model and a SUSY scenario. The event simulation starts from a proton proton collision leading to the process of interest all the way to the expected detector response.

These steps are the following :

- Event generation : The process of interest $pp \rightarrow X$ is generated by relying on random sampling using Monte Carlo (MC) techniques, which repeatedly draw samples that represent a possible outcome of a given process. The processes are generated using a software package, MADGRAPH [22] for example, which calculates the matrix element for each process to some order in QCD. Generators start from partons of a pp collision, using parton distribution functions, and calculate the processes up to leading order or next-to-leading order. For processes where the cross-section at a higher order is non-negligible a factor called the k -factor is applied to the expected cross-section. The partons from the hard interaction are coloured and radiate gluons described by a parton showering software, as PYTHIA [23]. The generators used in the analysis are given in Section 5.3. The raw output of such generators is an input to the next steps of simulation. They can also be used to perform generator level studies, also called “truth level”,

undergoing minimal processing to evaluate the sensitivity and acceptance of the analysis (for example Table 5.7).

- Detector simulation : The event generator gives particle momenta at the hadron level which are then processed by GEANT4 [24], which simulates the propagation of particles through the different materials comprising the detector. The simulation includes the best knowledge of the detector geometry, material budget and modeling of the particle interactions. The full simulation of the detector is a slow process. For many applications, such as the generation of SUSY signals, it is faster to use a parametrized response of the calorimeters [25].
- Digitisation : The detector simulation records the interaction of particles with the different components of the detector in the form of hits and energy deposits in the detector. The latter are used as inputs to emulate the response of the readout electronics of the detector. The output from this step is identical to the data recorded by the detector.
- Reconstruction : At this stage, both the recorded events by the detector and simulated events are used to identify objects associated with fundamental particles, namely electrons, muons, photons, and jets. The energy deposits not matched to physics objects are collected into a “soft terms” category used in the computation of the missing transverse momentum.

There are two other important elements of the simulation that are less understood. The simulated MC samples must handle the underlying event, which is the remainder of the non-hard scattered partons of the original interacting

protons. Also, the MC generators must simulate the interactions between the other particles in the beam crossing, also referred to as pileup. In practice, the MC is generated with an expected pileup profile which is latter corrected based on the observed pileup profile.

The final physics objects from the reconstruction step do not reflect all the knowledge we have about the detector. For instance, the energies of the objects must be calibrated or that certain parts of the detector may not always be working with the desired specification. Once the reconstructed objects are defined and calibrated, the data is ready for analysis. Typically, the size of the dataset is reduced from petabytes to a size of few terabytes by only selecting the objects of interest in the analysis. For instance, this analysis requires at least two leptons applied to the samples used which significantly reduce the sample size.

3.6 Reconstruction and Identification Techniques

The identification and reconstruction step relies on the properties that particles display when they interact with the different components of the detector. Once the particle is identified, it is desirable to determine its momentum, its origin, and its polarization among other properties. The important part for the analysis is how well these objects are reconstructed which can be determined by measuring the reconstruction and identification efficiencies as a function of kinematic particles, usually p_T and η . In the next sections we briefly describe how the main physics objects used in this analysis are identified and their efficiencies.

3.6.1 Electrons

Electrons are reconstructed and identified for $p_T > 5$ GeV in $|\eta| < 4.9$ range of the detector [26]. The electrons are identified by a cluster in the electromagnetic calorimeter of more than 2.5 GeV matched to a track in the inner detector. The electron trajectory is determined using information from the inner detector, like for the trajectory of any other charged particle. This involves the measurement of the track associated parameters : the position in the transverse (d_0) and longitudinal (z_0) planes of the perigee, the particle direction (ϕ, θ) and the parameter which provides the inverse track momentum multiplied by the charge (q/p). The track parameters and the associated uncertainties are obtained from the track fitting procedure performed with the ATLAS Global χ^2 Track Fitter.

The goal is to improve the efficiency of identifying electrons while rejecting background electrons arising from hadronic jets mistaken for electrons, electrons from photon conversion, Dalitz decays and from semileptonic heavy-flavour hadron decays. To do so, three identification (ID) criteria, loose, medium, and tight, are defined via likelihoods based on calorimetric cluster shower shapes, track and track-to-cluster matching variables. The tighter the ID criteria, the higher the rejection of background electrons but the lower the identification efficiency. Figure 3.13 shows the efficiencies in data and MC for three operating points that are based on a likelihood approach, Loose, Medium and Tight. The data efficiencies are obtained by applying data/MC efficiency ratios that were measured in $J/\psi \rightarrow e^+e^-$ and $Z \rightarrow e^+e^-$ events to MC simulation. The lower efficiency in data than in MC

arises from the fact that the MC does not properly represent the 2016 TRT conditions, in addition to the known mismodelling of calorimeter shower shapes in the GEANT4 detector simulation. The reconstruction efficiency of electrons is around 95% for low p_T and goes up to 99.9% for electron $p_T > 45$ GeV.

To study and compute the corrections needed to account for the detector geometry and material distribution two sets of MC samples are used. For the first set an ideal geometry (no misalignments) with the best knowledge of the dead material is implemented. For the second scenario, the dead material between the tracker and calorimeters is increased and the misalignments are included. The latter is used to assign the systematic uncertainties.

3.6.2 Muons

The ATLAS detector has been designed to provide clean and efficient muon identification and precise muon momentum measurements over a wide range of energy and solid angle. The muon reconstruction starts in the inner detector where tracks are identified up to $|\eta| < 2.5$ in a solenoidal field of 2 Tesla . The muon spectrometer measures muons up to $|\eta| < 2.7$ providing momentum measurements with a design relative resolution of better than 3% over a wide p_T range and to 10% at TeV. Similar to electrons, four muon identification selections are defined in order to meet the specific needs of different physics analyses :

- loose muons : maximize efficiency : ideal for multilepton final states analysis
- medium muons : minimize systematics uncertainties
- tight muons : optimize purity

- high- p_T muons : maximize momentum resolution for high p_T tracks (> 100 GeV)

Figure 3.14 shows the reconstruction efficiency of the muon for different selections measured in $Z \rightarrow \mu\mu$ and $J/\psi \rightarrow \mu\mu$ events.

The reconstruction efficiency is measured to be close to 99% over most of the phase space relevant for the analysis. The isolation efficiency is between 93% and 100% based on the selection and muon momentum. The simulation reproduces both efficiency very well. The momentum resolution is measured to be as low as 1.7% and the momentum scale uncertainty is less than 0.05%.

3.6.3 Jets

Jets are reconstructed using the anti- k_t jet algorithm [27] with the distance parameter R set to 0.4 and a three dimensional input of topological energy clusters in the calorimeter [28]. The advantage of using the anti- k_t algorithm is that it is infrared and collinear safe. It is also resilient to soft-QCD emissions, a process that is common in the hadron colliders. The jets are constructed by defining two distances :

- $d_{ij} = \min((k_{kj}^{2p}, k_{kj}^{2p}) \frac{\Delta_{ij}^2}{R^2})$: the distance between two particles i and j.
- $d_{iB} = k_{ti}^{2p}$: the distance between a particle i and the beam B .

where $\Delta_{ij}^2 = (y_i - y_j)^2 + (\phi_i - \phi_j)^2$ and k_{ti} , y_i , and ϕ_i are the transverse momentum, rapidity, and azimuth of the particle i . The radius parameter R scales d_{ij} with respect to d_{iB} such that any pair of final jets a and b are separated by at least $R^2 = \Delta_{ab}^2$. The parameter p governs the relative power of energy with respect

to the geometrical scales, and is set to $p = -1$ for the anti- k_t algorithm.

The energies of the jets are calibrated to account the necessary losses associated with sampling calorimeter, the presence of dead material, energy loss in non-instrumented regions, etc. This is performed using the local cluster weighting (LCW) scheme [29], which uses calibrated toplogical clusters as input the anti- k_t jet algorithm, and taking into account jet energy scale (JES) and jet energy resolution (JER) calibrations. Both JES and JER can have an important contribution to the systematic uncertainties of this analysis. The fractional JES uncertainty in Figure 3.15 shows that jets with p_T below 50GeV can have a larger uncertainty. The choice was made to only require jets above 50 GeV in the analysis for this reason. The fractional JER is around 17% for jets with p_T of 30 GeV decreasing down to 5% for more energetic jets. An additional variable is used to suppress jets from pileup, called the jet vertex tagger (JVT). JVT is a multivariate combination of the fraction of the total momentum of tracks in the jet which is associated with the primary vertex and track-based variables to suppress pileup jets [30].

3.6.4 Heavy flavor

Jets containing b -hadrons (commonly referred to as b -tagging) is performed with the MV2 algorithm, a multivariate discriminant making use of track impact parameters and reconstructed secondary vertices [70, 71] to provide the best separation among the different jet flavour hypotheses. An example of the output of the multivariate discriminant is shown in Figure 3.16a where typically a cut must be applied on the score to identify an operating working point. The notation

MV2cxx denotes the fraction xx of c-jet contribution present in the training. For instance, MV2c00 denotes the MV2 algorithm where no c-jet contribution was present in the training and MV2c10 (MV2c20) denote the MV2 outputs where a 7% (15%) c-jet fractions was present in the background sample. The performance of the optimized MV2c00, MV2c10 and MV2c20 b-tagging algorithms is shown in Figure 3.16b for the c-jet rejection as a function of the b-jet efficiency. The MV2c20 is the best performing algorithm leading to an optimal rejection of c-jets at a given b-tagging efficiency. As a result, it is the algorithm used in this analysis.

3.6.5 Missing transverse momentum

The missing transverse momentum is an important quantity in searching for new physics scenarios which expect a stable and non-electromagnetically interacting particle. Such particle does not interact with the detector and can only be identified through an imbalance of the momentum distribution between the particles of the event. Since the momentum of the colliding protons is almost along the beam, longitudinal, the transverse component of the momenta of the scattered objects should add up to zero. Based on the conservation of momentum, the sum of all visible four momenta projected in the transverse plane should be close to zero if no particles are missed. However, this quantity will be large if a particle, potentially from new physics models, escaped detection. The missing transverse momentum is defined as the negative vector sum of the transverse momenta of the visible

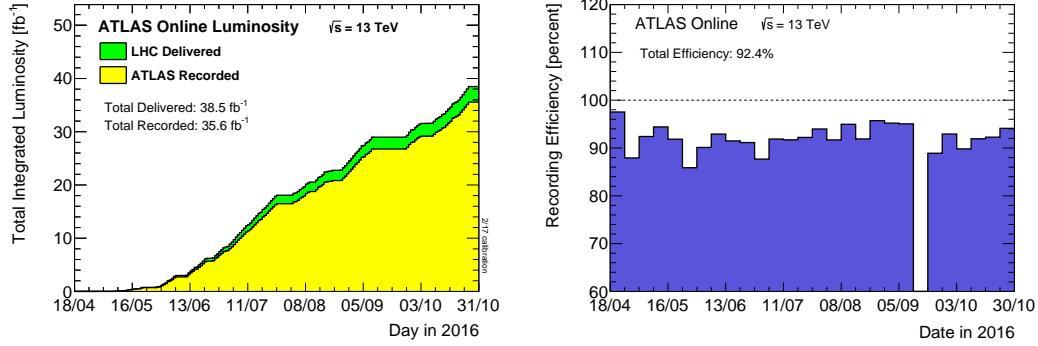
reconstructed objects in the event :

$$p_T^{\text{miss}} = - \sum_{\text{visible}} p_T \quad (3.5)$$

where the visible objects include electrons, muons, jets, photons, taus, and a soft term. In the rest of the dissertation, the magnitude of the missing transverse momentum vector is denoted by E_T^{miss} . The soft term is a fundamental quantity in the reconstruction of E_T^{miss} and be estimated by

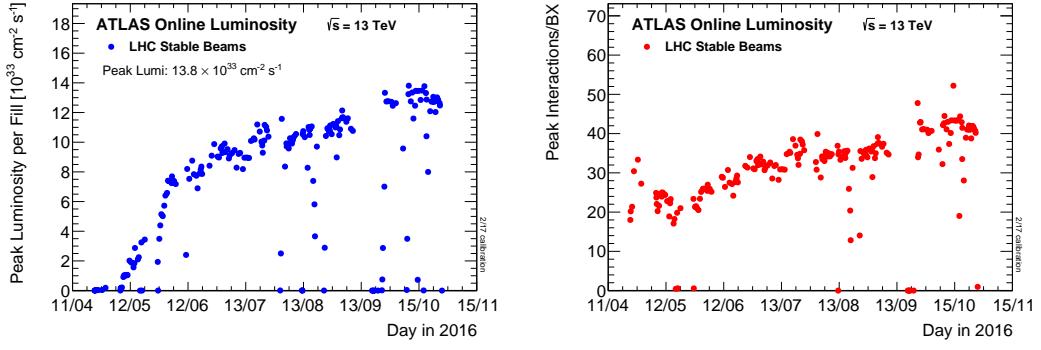
- Calorimeter based Soft Term (CST) : accounts for both neutral and charged particle energies.
- Track based Soft Term (TST) : incorporates a natural pileup suppression by selecting only tracks from primary vertices.

The E_T^{miss} performance depends on the event topology affected by the presence of true E_T^{miss} , from neutrinos for example, charged leptons, jet activity, and others. The E_T^{miss} performance is generally studied with processes with and without genuine E_T^{miss} , such as $W \rightarrow e\nu$ and $Z \rightarrow \mu\mu$ events. The scale and resolution for the reconstructed E_T^{miss} in these processes are indicative of the reconstruction quality. For illustration, results obtained with $Z \rightarrow \mu\mu$ events are shown in Figure 3.17. Generally, the E_T^{miss} has a resolution in the order of 10 to 20 GeV.



(a) Total Integrated Luminosity in 2016 (b) Data Taking Efficiency per Week in 2016

FIGURE 3.4: (a) Cumulative luminosity versus time delivered to (green) and recorded by ATLAS (yellow) during stable beams for pp collisions at 13 TeV centre-of-mass energy in 2016. The recorded luminosity reflects the DAQ inefficiency, as well as the inefficiency of the so-called “warm start” : when the stable beam flag is raised, the tracking detectors undergo a ramp of the high-voltage and, for the pixel system, turning on the preamplifiers. (b) ATLAS data taking efficiency in 2016. The denominator is the luminosity delivered between the declaration of stable beams and the LHC request to turn the sensitive detectors off to allow a beam dump or beam studies. The numerator is the luminosity recorded by ATLAS. Each bin represents one day. The empty bins are due to days in which no stable beams were delivered by the LHC. The inefficiency accounts for the time needed to turn on the high-voltage of the Pixel, SCT, and some of the muon detectors at the start of an LHC fill and any inefficiencies due to deadtime or due to individual problems with a given sub detector that prevents the ATLAS data taking to proceed.



(a) Peak Luminosity per Fill in 2016 (b) Peak Interactions per Crossing in 2016

FIGURE 3.5: (a) The peak instantaneous luminosity delivered to ATLAS during stable beams for pp collisions at 13 TeV centre-of-mass energy is shown for each LHC fill as a function of time in 2016. The luminosity is determined using counting rates measured by the luminosity detectors, and is based on a preliminary analysis of van-der-Meer beam-separation scans during 2016. (b) The maximum number of inelastic collisions per beam crossing (μ) during stable beams for pp collisions at 13 TeV centre-of-mass energy is shown for each fill in 2016. The preliminary luminosity measurement is used to determine the number of interactions per beam crossing as $mu = L_{\text{bunch}}\sigma_{\text{inel}}/f_r$ where L_{bunch} is the per-bunch instantaneous luminosity, σ_{inel} is the inelastic cross-section at 13 TeV, which is taken to be 80 mb, and f_r is the LHC revolution frequency of 11.245 kHz. The number of interactions shown is averaged over all colliding bunch pairs, and only the peak value per fill during stable beams is shown.

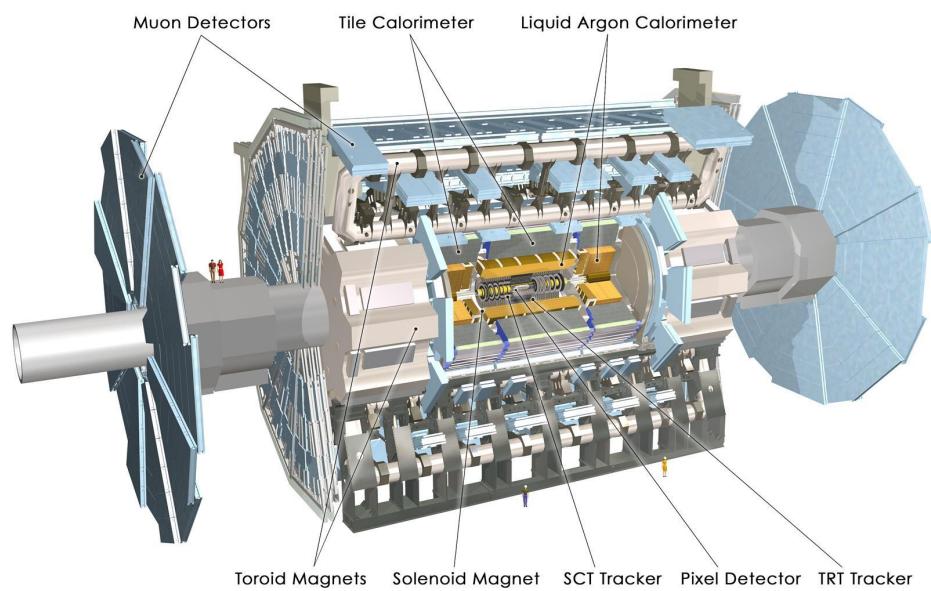


FIGURE 3.6

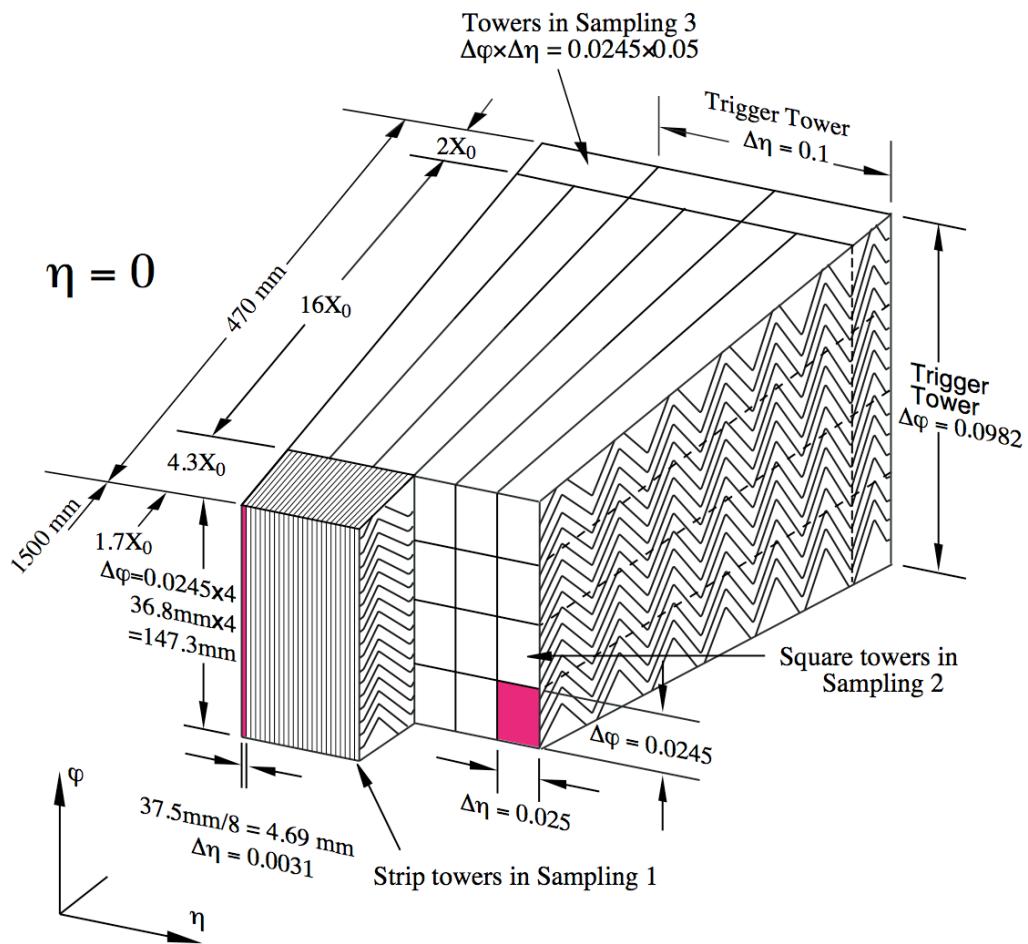


FIGURE 3.7: Sketch of the accordion structure of the EM calorimeter.

FIGURE 3.8

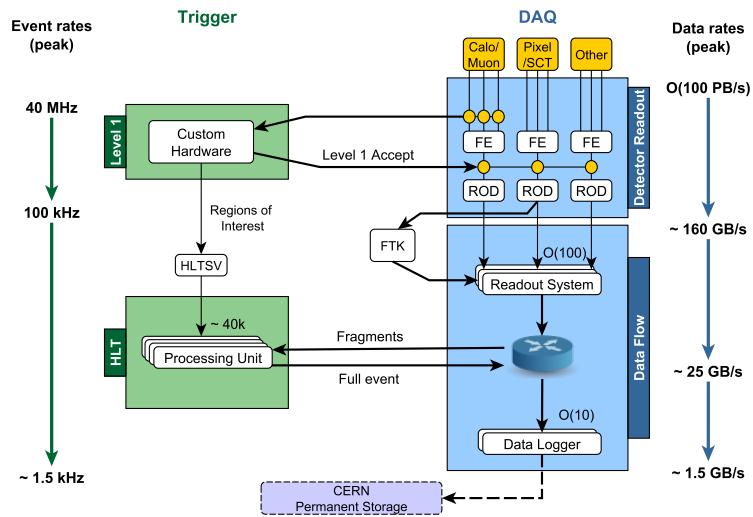


FIGURE 3.9: ATLAS TDAQ architecture.

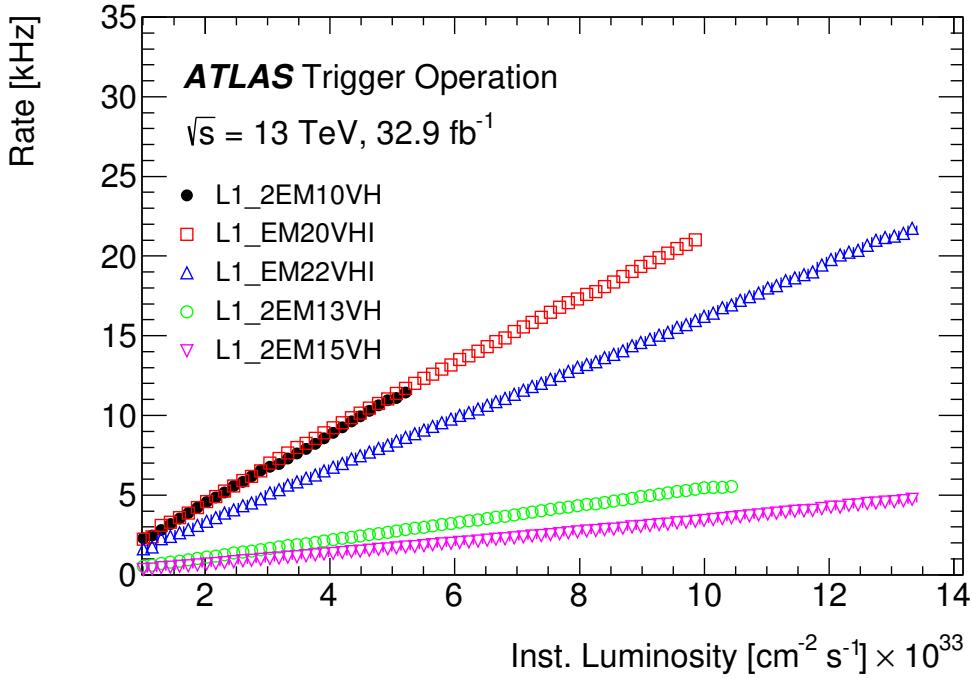


FIGURE 3.10: Output rates of Level-1 EM triggers as a function of the uncalibrated instantaneous luminosity measured online during the 2016 proton-proton data taking at a center-of-mass energy of 13 TeV. Rates are shown only for unprescaled triggers. All trigger rates show a linear dependency with instantaneous luminosity.

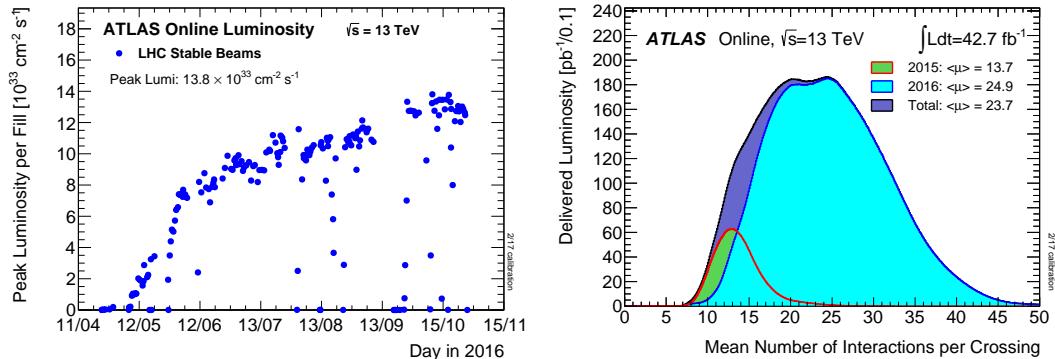


FIGURE 3.11: Run conditions during Run 2 : ATLAS peak online luminosity per fill (left), ATLAS online pileup (right) [18].

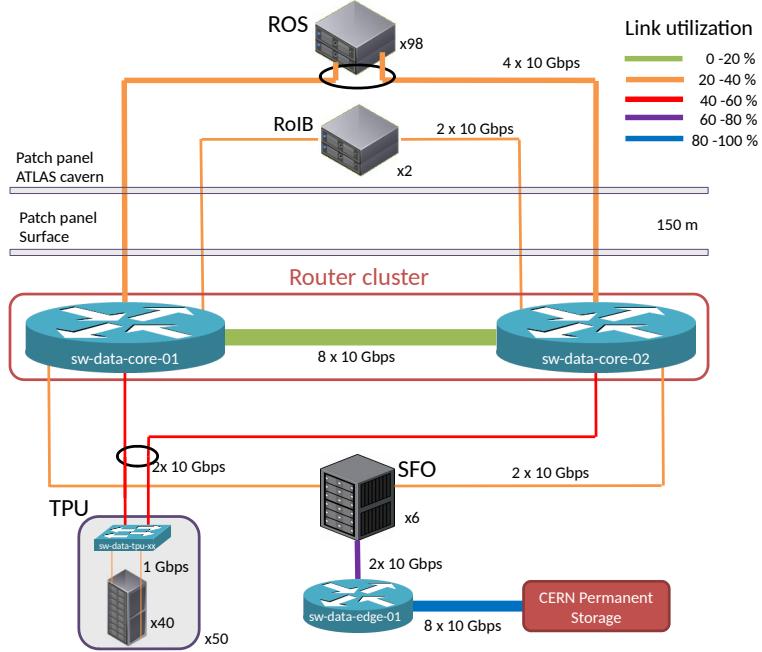


FIGURE 3.12: ATLAS Dataflow network

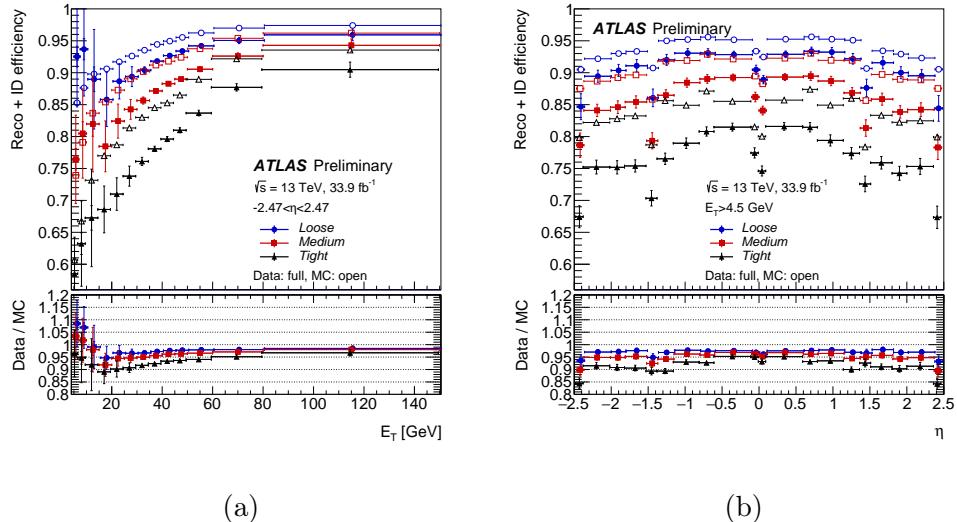


FIGURE 3.13: Electron reconstruction and identification efficiencies in $Z \rightarrow e^+e^-$ events as a function of (a) pseudo-rapidity η (b) transverse energy E_T . The total statistical and systematic uncertainty is displayed.

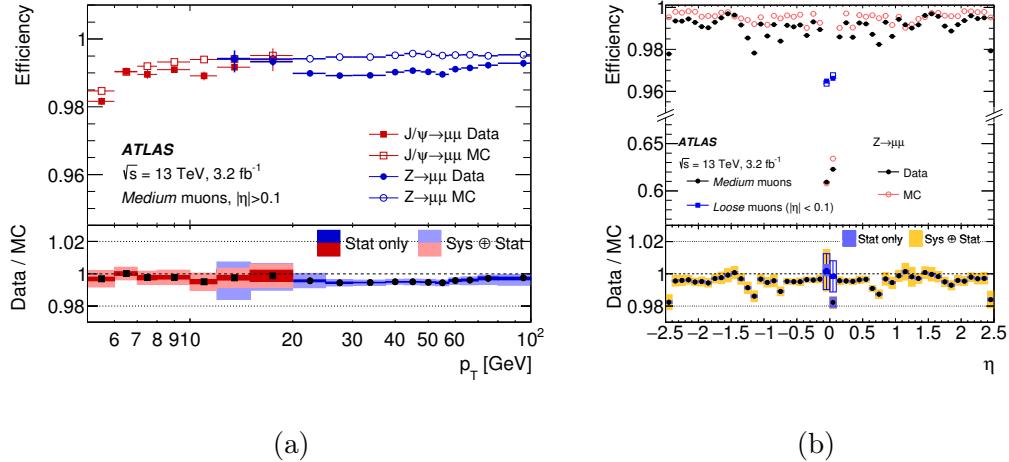


FIGURE 3.14: (a) Reconstruction efficiency for the medium muon selection as a function of (a) p_T (b) η of the muon. The error bars on the efficiencies indicate the statistical uncertainty. The panel at the bottom shows the ratio of the measured to predicted efficiencies, with statistical and systematic uncertainties.

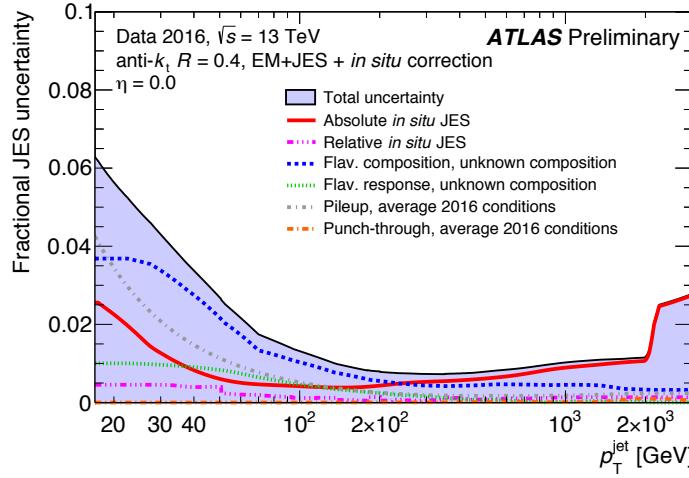


FIGURE 3.15: Fractional jet energy scale systematic uncertainty components as a function of p_T for anti- k_t jets at $\eta = 0.0$ with distance parameter of $R = 0.4$ after calibration. The total uncertainty (all components summed in quadrature) is shown as a filled blue region topped by a solid black line.

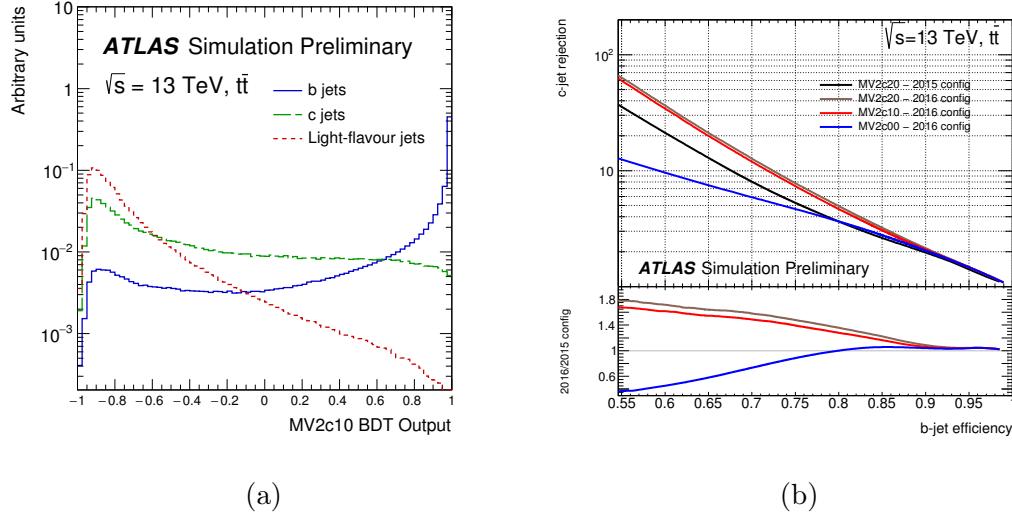


FIGURE 3.16: (a) MV2c10 BDT output for b- (solid blue), c- (dashed green) and light-flavour (dotted red) jets evaluated with $t\bar{t}$ events. (b) c-jet rejection versus b-jet efficiency for the 2015 and 2016 configurations of the algorithm. (for the 2016 configuration).

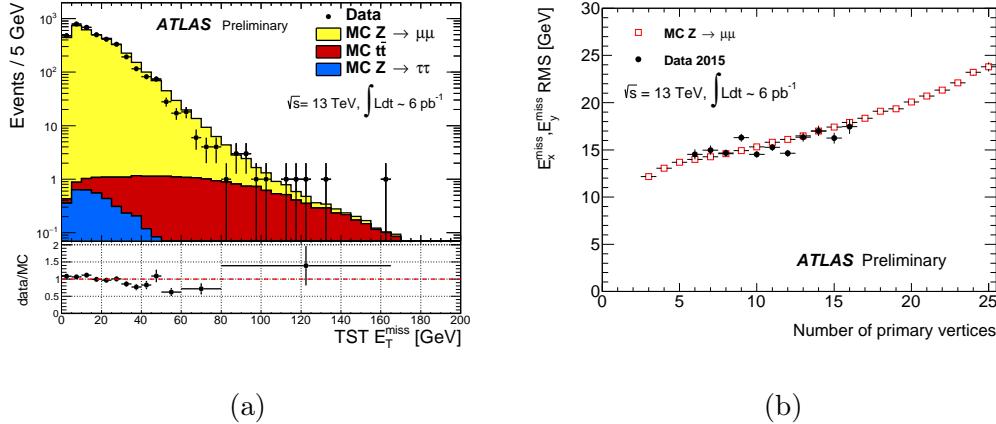


FIGURE 3.17: (a) E_T^{miss} distribution as measured in data with $Z \rightarrow \mu\mu$ events without pile-up suppression. (b) E_T^{miss} resolution as a function of the number of primary vertices in $Z \rightarrow \mu\mu$ events. The data (black circles) and MC simulation (red squares) are overlaid.

Chapitre 4

The Region of Interest Builder

4.1 Overview

The TDAQ system reduces the proton interaction rate from 40 MHz to the ATLAS data storage capacity of about 1.5 kHz. A hardware First Level Trigger (L1) reduces the rate to 100 kHz and a software High Level Trigger (HLT) selects events for offline analysis. Jet, electromagnetic and tau clusters, missing transverse momentum (E_T^{miss}), $\sum E_T$, jet E_T , and muon candidate information from L1 determine detector Regions of Interest (RoIs) that seed HLT processing. These RoIs are provided to the HLT by a custom VMEbus based system, referred to as the Region of Interest Builder (RoIB) [31]. The RoIB collects data from L1 trigger sources and assembles the data fragments into a complete record of L1 RoIs. These RoIs are made available to the HLT to initiate event processing. In order to improve maintainability and scalability, and to minimize the amount of custom hardware needing to be supported, the RoIB was implemented using commodity server hardware and an interface technology deployed within the ATLAS Trigger and Data Acquisition (TDAQ) system. The approach of implementing the RoIB functionality in software has been investigated in the past and the conclusion at that time was that a software based approach is possible but requires a card with a higher readout rate [32]. Since data readout cards operating at high rates became available and the capabilities of computers have improved with the increase in

CPU clock speed and number of cores, it became possible to implement the RoIB functionality using a PC based approach. The PC based RoIB must duplicate the functionality of the VMEbus based RoIB which means that the PC based solution must receive and assemble the individual L1 fragments, and pass them as a single L1 result to the HLT. Modern computers have multicore CPU architectures with the possibility of running multi-threaded application, a feature which is being fully exploited in the RoIB software to achieve the desired performance of 100 kHz over 12 input links for fragment sizes of 400 bytes. This chapter describes the work of the author in evolving the RoIB from the VMEbus based system to the PC based system and gives details on the hardware, firmware, and software designs used to achieve the full RoIB functionality.

4.2 VMEbus based RoIB

4.2.1 Hardware implementation

The RoIB was implemented as a custom 9U VMEbus system that includes a controller which configures and monitors the system along with custom cards that receive and assemble the event fragments and send them to the HLT. Figure 4.1 shows a block of the RoIB and its connection to external systems.

The RoIB contains four input cards and uses one builder card in the Run-2 configuration. Each input card accepts three inputs from L1 subsystems. The builder card assembles the input data of the events and passes the results via two optical links to another receiver card in a PC running the HLT supervisor

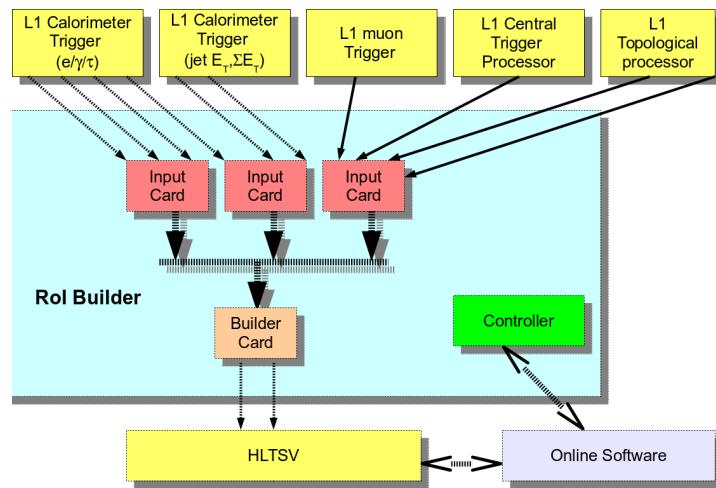


FIGURE 4.1: Block scheme of the RoI Builder and overview of connections to external systems. The custom input and builder cards and the controller, a commercially available single board computer, are installed in a single 9U VMEbus crate. The controller connects to the Control Network to interact with the rest of the data acquisition system.

(HLTSV) application. The receiver card in the HLTSV is a TILAR card [33] that implements four PCIe Gen1 lanes to interface with the two optical links. The HLTSV manages the HLT processing farm by using L1 results provided by the RoIB, retrieves events from the ROS, assigns events to HLT farm nodes, and handles event bookkeeping including requesting removal of data from ROS storage when no longer required.

The fragments received by the RoIB are identified by a 32 bit identifier, the extended L1 ID (L1ID). The RoIB input cards use the L1ID and the number of outputs enabled to assign keys to the various fragments and send them to the output channel in the builder card that was assigned that key value. The input data is transferred over a custom J3 back-plane. The back-plane operates at 20 MHz and transfers 16 data bits per clock cycle simultaneously for up to 12 inputs. The total maximum data throughput is therefore 480 MB/s, 40 MB/s per input. The maximum size of any single fragment is limited to 512 bytes imposed by resources available in the FPGA firmware. The current RoIB input links are listed in Table 4.1.

4.2.2 System Performance and Evolution

The custom VMEbus based RoIB operated reliably during the first run of the LHC, however, it is desirable to have a more flexible RoIB. In addition, the RoIB is getting close to its design limitation, as seen in Figure 4.5. For fragments of 400 bytes and inputs from eight L1 systems, referred to as channels, the current RoIB rate limit is 60 kHz which is below the required 100 kHz at L1. While the current

TABLE 4.1: L1 input sources to the RoIB.

Source	Links
Central Trigger Processor (CTP)	1
L1 calorimeters ($e/\gamma, \tau, \text{jet}, \sum E_T$)	6
Muon Trigger to CTP Interface (MUCTPI)	1
Topological processor (L1Topo)	2
Spare	2

fragment size coming from L1 are around 160 bytes, the sizes are expected to grow due to the increase of instantaneous luminosity and the complexity of L1 triggers. The current VMEbus system will be replaced by a PCI-express card hosted in the HLTSV PC with the possibility to upgrade the commodity hardware (e.g. ability to upgrade CPUs). The new configuration simplifies the readout architecture of ATLAS. The targeted rate for event building is 100 kHz over 12 input channels for fragment sizes in the order of 400 bytes.

4.3 PC based RoIB

A custom PCIe card developed by the ALICE collaboration, the Common ReadOut Receiver Card (C-RORC) [34], was deployed as an upgraded detector readout interface within the ATLAS ROS with ATLAS specific firmware and software called the RobinNP [35]. The new PC based RoIB uses the RobinNP firmware and a dedicated API to facilitate the implementation of the RoIB

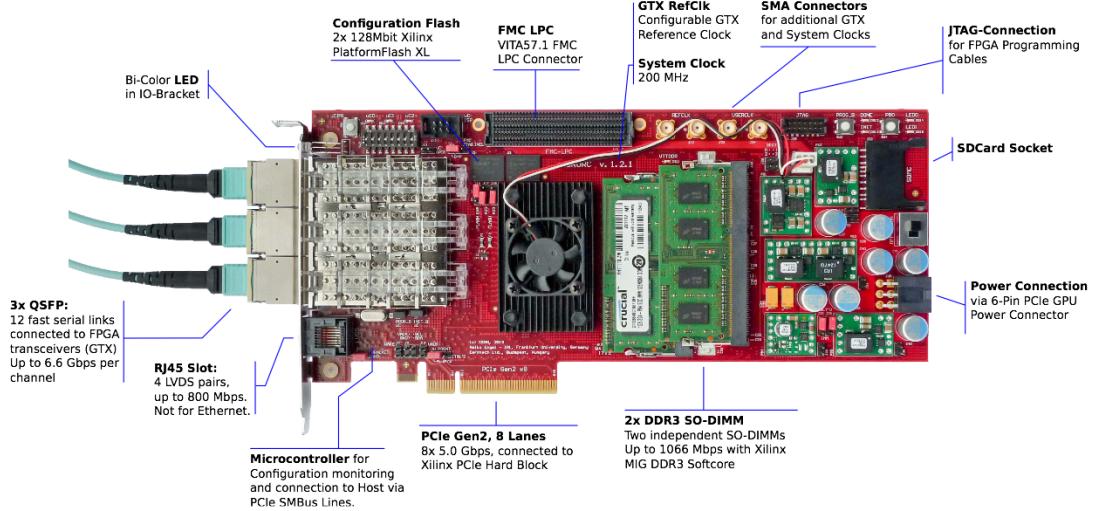


FIGURE 4.2: Photo of the C-RORC board with the major components and features annotated [35].

functionality on a commodity PC. In this section, we describe the C-RORC hardware as well as the RobinNP firmware, API, and the event building software.

4.3.1 The Common Readout Receiver Card

The C-RORC implements 8 PCIe Gen1 lanes with 1.4 GB/s bandwidth to the CPU fed via 12 optical links each running 200 MB/s on 3 QSFP transceivers. It utilizes a single Xilinx Virtex-6 series FPGA that handles data input from the 12 links and buffers the data in two on-board DDR3 memories. It is also capable of processing and initiating DMA transfer of event data from the on-board memory to its host PC's memory. The major components of the C-RORC are annotated in the picture shown in Figure 4.2.

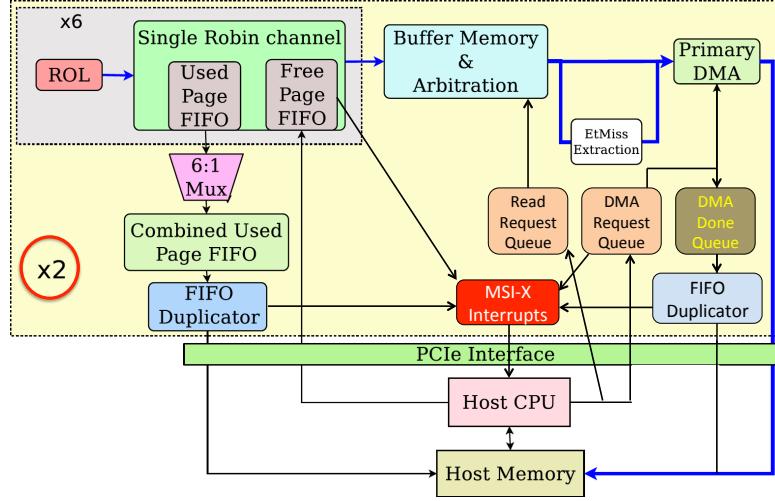


FIGURE 4.3: RobinNP firmware organization and flow of data from host CPU to the firmware (by means of programmed I/O) and from the firmware to the host memory (by means of DMA).

4.3.2 Readout System Firmware and Software

The RobinNP firmware used for the RoIB is identical to that used in the ATLAS ROS[17]. As shown in the schematic of Figure 4.3, the logic is divided into two functional blocks, known as sub-ROBs, each servicing six input links and one DDR3 memory module. Event data fragments arriving via a link are subjected to a range of error checks before being stored in the memory module for the relevant sub-ROB. At the same time a token representing the address of a region of the memory, referred to as a page, is passed to a listening software process via a ‘FIFO duplicator’. To avoid a costly read across the PCIe bus, data is continuously streamed from firmware to software via a chain of firmware and software FIFOs. Notification of new data arriving in the software FIFO is

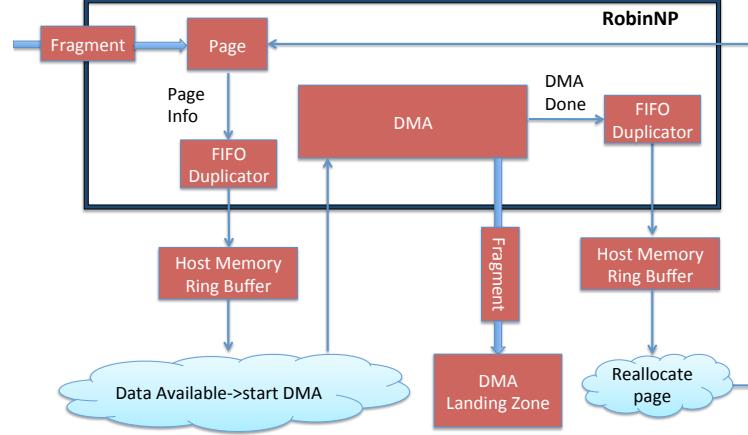


FIGURE 4.4: Layout of the readout system firmware and software specific to the RoIB.

managed via coalesced interrupts to allow for efficient use of CPU resources. For the RoIB application, the receipt of page information immediately triggers a DMA of fragment data from the RobinNP memory into the host PC memory. The fragments are then passed via a queue (one per sub-ROB) to the RoIB process along with any relevant fragment error information. A schematic of this shortened dataflow path is presented in Figure 4.4. The API for the RoIB process consists of these queues, return queues for processed pages now available for re-use and a configuration interface. The software is implemented with multiple threads each handling specific tasks such as supply of free pages, receipt of used pages, DMA control and bulk receipt of fragment data.

4.3.3 RoIB Software

The HLTSV is a multi-threaded application that obtains a L1 result from a variety of possible input sources and exchanges information with the rest of the

HLT computing farm. For the RoIB, the L1 source is a RobinNP interface that performs fragment assembly and is used as a plug-in to the HLTSV application. The RobinNP plug-in has two receive threads, each thread services six channels by pulling fragments from the RobinNP on-board memories to the host PC. Fragments with the same L1ID are copied to a contiguous memory space and a queue of completed events is prepared. Upon request by the HLTSV, a pointer to the contiguous memory space is passed back to the HLTSV process for further handling. In order to optimize concurrent access to RoIB data structures, containers from the Intel threading building block (TBB) library were used. These containers allow multiple threads to concurrently access and update items in the container while maintaining high performance.

4.4 Prototype Tests

In order to understand the requirements for the underlying server PC, a validation system based on Intel(R) Xeon(R) CPU E5-1650 v2 @ 3.5 GHz with six cores is being used to perform tests of the PC based RoIB. The goal is to perform software based fragment assembly at a rate of 100 kHz over 12 channels for a typical fragment size of 400 bytes. The current system offers flexibility in terms of the fragment size allowed which was not the case in the VMEbus based RoIB. The initial tests were performed with a standalone application that implements a minimal interface for event building. Once the system was validated, the relevant code modules were integrated into an HLTSV process running within the full

ATLAS TDAQ software suite with appropriately scaled test hardware to represent the remaining elements of the system.

4.4.1 Standalone Tests

The goal was to test input/output bandwidth limitations of the RobinNP and the rate of event building. Initial performance testing used a standalone RobinNP application and an external source that emulates the L1 trigger data in the form of 32-bit word fragments with 12 channels. In this test, the host PC was running the assembly routine with a single threaded application. Figure 4.5 shows the input rate without event building as a function of fragment size. For 400 byte fragments the input rate to the RobinNP is 215 kHz. The same figure shows the event building rate which is 150 kHz. This performance shows that the event building at the required rate of 100 kHz with 12 channels is achievable in a standalone application.

4.4.2 Full System Tests

Since the HLTSV is performing tasks other than the event building, there is overhead associated with additional operations that reduces the performance. For this reason, we use the full ATLAS TDAQ software in a test environment that emulates the major components of the ATLAS data acquisition system shown in Figure 3.9. The setup includes an emulated input from L1 trigger sources, the HLTSV and other PCs to simulate the HLT computing farm, and the ROS that buffers the full event data. In this test setup, an external source sends data

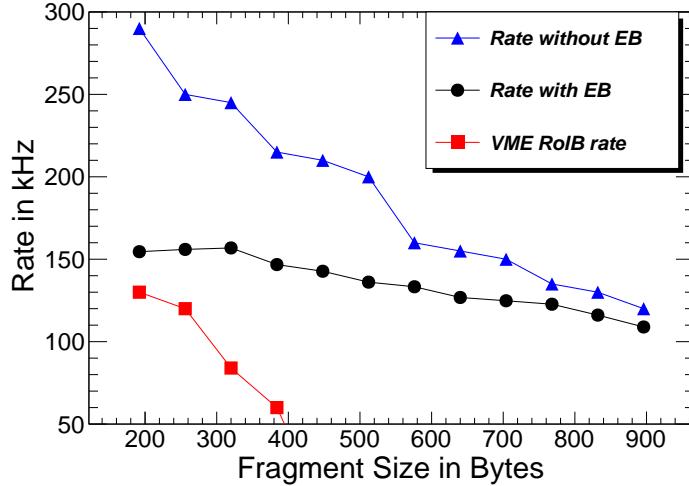


FIGURE 4.5: Rate as a function of the fragment size (in bytes) with external source that emulates the L1 trigger input. The rates shown are for the input rate to the RobinNP without event building (EB) (triangle), rate with EB (circle), and for comparison, the current VMEbus RoIB rate is also shown (square).

that emulates L1 RoIs via 12 links connected to the RobinNP hosted by the HLTSV. When the HLTSV requests a built RoI event, the software RoIB plug-in provides the RoI event which will be used to seed requests for the event data to be processed. Figure 4.6 shows an event building rate of 110 kHz measured with 400 byte fragments with the HLTSV application in a setup close to the ATLAS TDAQ system.

As shown in Figure 4.7, the performance of the PC-RoIB with realistic running ATLAS conditions is improved over the VME-RoIB particularly at high RoI sizes and maintains a rate of over 100 kHz with 12 channels.

The design specification of the ATLAS L1 trigger is to send data at 100 kHz. While the tests above showed that the PC RoIB meets the desired rate requirement

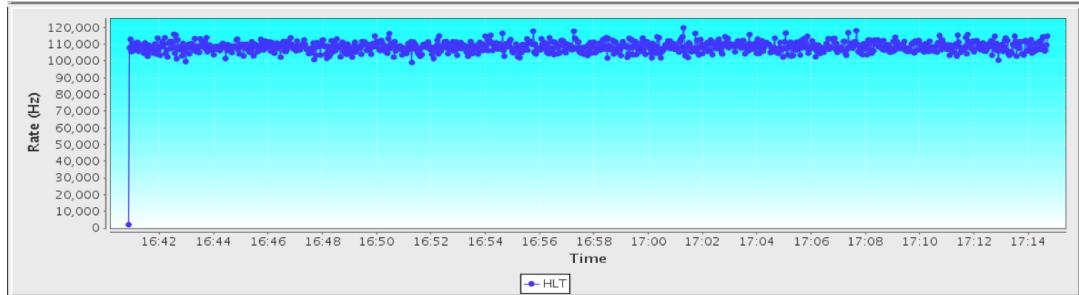


FIGURE 4.6: Screenshot of a monitoring tool which shows the HLTSV processing rate using the ATLAS TDAQ software.

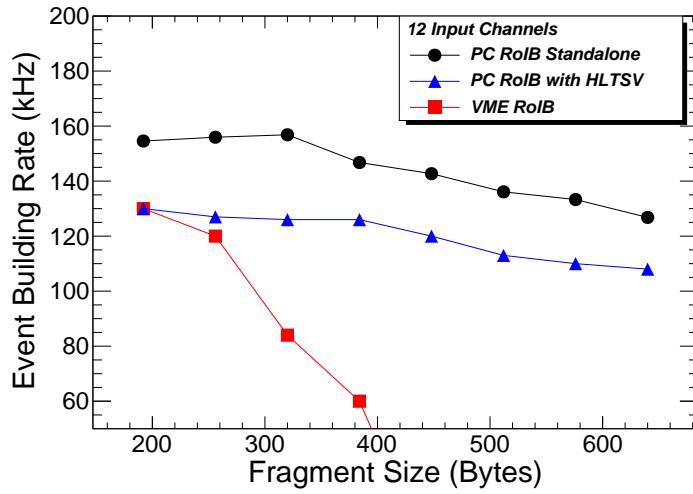


FIGURE 4.7: The event building rate as a function of the RoI record size in Bytes. The rates are shown for a standalone application that implements a minimal interface for event building, the integrated RoIB software into an HLTSV process running within the full ATLAS TDAQ software suite, and for comparison the VME-RoIB performance.

in the case that an external source is sending data as fast as possible (much more than 100 kHz), it is important to test that the PC RoIB will sustain the 100 kHz rate if the external source sends data at exactly 100 kHz. Figure 4.8a demonstrates

that in the event that the incoming data to the PC RoIB is fixed at 100 kHz, the event building in the PC RoIB still operates at this rate. The other important variable that can affect the rate is the number of channels. In particular, the ATLAS detector might decide to disable some of the channels which should not affect the rate of operation of the PC RoIB. Figure 4.8b shows that the PC RoIB will operate at even higher rates if the number of channels is reduced.

With these tests, the author validated the operation of the new PC RoIB which deemed it ready to be deployed in the ATLAS system.

4.5 Online Performance in Run-2

The author deployed the new PC RoIB in the ATLAS trigger and data acquisition system during the LHC winter shutdown in January 2016. Initially, the PC RoIB was used as the main system with the VMEbus RoIB was used as a backup system. Later, the VMEbus RoIB was removed completely and the PC RoIB became the only system running in the ATLAS trigger. The PC RoIB operated reliably since its installation without any problems and without deadtime for the ATLAS data collection. Figure 4.9a shows that the RoIB event assembly does not depend on pileup conditions and Figure 4.9b shows that the memory usage of the HLTSV is at the level of 5%. It has now participated in collecting a dataset of over 35 fb^{-1} exceeding the dataset collected by the VMEbus RoIB (22 fb^{-1}). The performance of the PC RoIB during the data taking of ATLAS has been very stable.

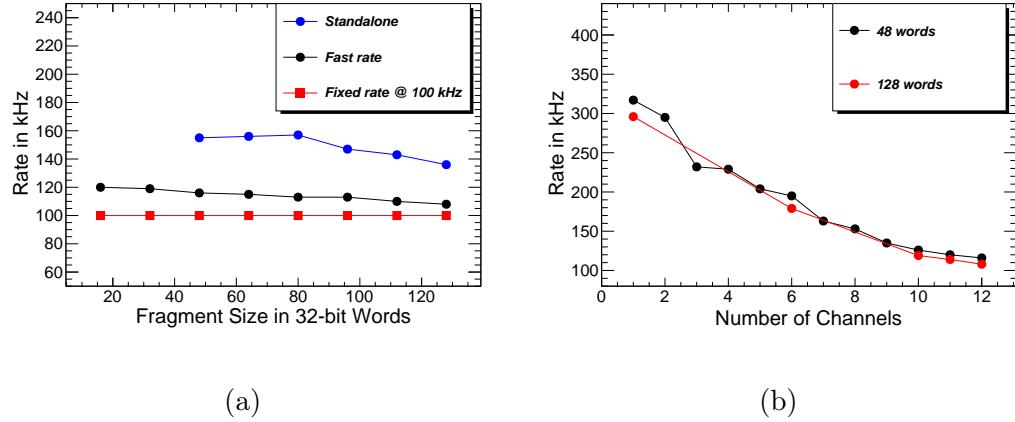


FIGURE 4.8: The event building rate as a function of (a) the fragment size (b) the number of channels. The rates are shown for a standalone application that implements a minimal interface for event building, the integrated RoIB software into an HLTSV process running within the full ATLAS TDAQ software suite.

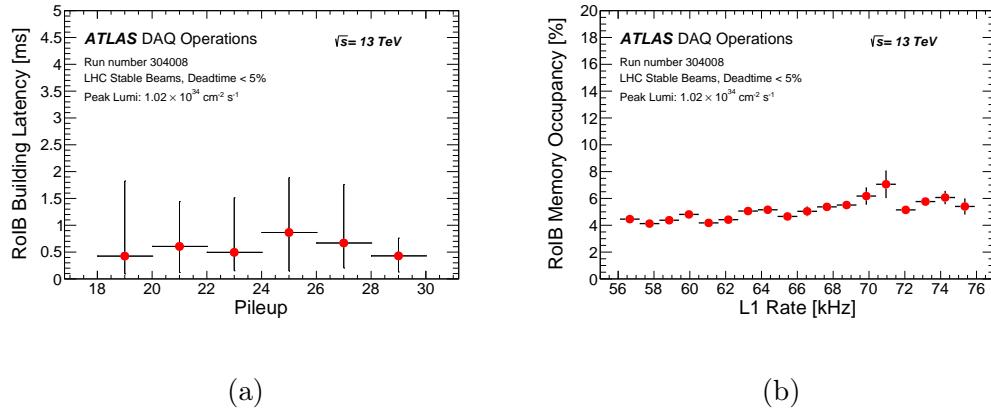


FIGURE 4.9: RoIB performance : RoIB building latency as a function of pileup (left), RoIB memory occupancy as a function of L1 rate (right).

Chapitre 5

Analysis Strategy

5.1 Overview

The chapters thus far described supersymmetry as an extension of the standard model, motivated the need to search for it, and described the experimental apparatus used for this search, the LHC and ATLAS. At this stage, we are ready to discuss the search for supersymmetry that the author has performed.

The main task in designing a search for new physics signatures is to ensure that the search regions, expected to have an enhancement of the signal (also referred to as the signal regions), have an acceptance for a wide range of new physics models that are well motivated. In addition, it is important to design the search in a way that minimizes the contamination from the known physics processes of the standard model. In other words, a typical signal region should have a maximum expected signal with a minimum expected background.

The search for supersymmetry with leptons of the same-electric charge or more than three leptons meets both criteria. The search targets the strong production of supersymmetric particles which mainly involves gluino pair production. Since gluinos are Majorana fermions, they can decay to either a positive or negative lepton in each branch of the pair production. As a result, it is as likely to have two leptons of the same electric charge as it is to have two leptons of opposite electric charge. While the opposite-sign leptons is a common signature in the standard

model, the same-sign leptons signature is very rare. Also, processes that involve more than two leptons are also rare in the standard model. On one hand the electroweak processes leading to W and Z bosons have a low cross sections. On the other hand the low branching ratio to leptons lead to an extreme background reduction. The requirement of three or more leptons allows the analysis to target supersymmetric models with longer decay chains. The presence of a third softer lepton also increases the sensitivity to scenarios with small mass difference between the supersymmetric particles.

The next important step in a search for new physics is to estimate the backgrounds present in the signal regions. The standard model backgrounds with a same-sign lepton pair or three or more leptons predominantly comes from the associated production of a top quark pair and a vector boson ($t\bar{t} + W$, $t\bar{t} + Z$), and multi-boson production (di-boson and tri-boson). These backgrounds that lead to a signature with same-sign leptons or three or more leptons are referred to as *irreducible backgrounds*. However, the high cross section processes from the standard model such as $t\bar{t}$, might contribute to the signal regions via a mis-reconstruction of this process by the detector. As a result, there are two very important backgrounds that affect the analysis that are referred to as *reducible backgrounds*. The top quark pair production ($t\bar{t}$) process may decay fully leptonically ($t\bar{t} \rightarrow (b\ell^+\bar{\nu})(\bar{b}\ell^-\nu)$) and contribute to the signal regions if the lepton charge is mis-measured. In the case of a semi-leptonic decay of $t\bar{t}$ ($t\bar{t} \rightarrow (b\ell^+\bar{\nu})(\bar{b}q\bar{q}')$) where the hadronic decay is mis-identified as a leptonic decay, the process will contribute with a “fake” lepton in the signal regions. The background estimation

methodology will aim at estimating both reducible and irreducible backgrounds with Monte Carlo simulation and data-driven methods.

Finally, we assess the compatibility between the observed data and the predicted background in one counting experiment for each signal region by doing a hypothesis test of the background-only or the background as well as the sought after signal hypotheses. If an excess is found in data, we proceed to evaluate if this excess can lead to reject the background-only hypothesis and we check the plausibility that the new signal can describe the data. Otherwise, we set exclusion limits for a certain region of the parameter space of a defined model or we set model-independent upper limits on the number of events from a beyond the standard model process.

5.2 Phenomenology (Benchmarking Models)

Final states with two same-sign leptons or three leptons and multiple jets can probe a variety of supersymmetric models represented by decays of heavy superpartners involving massive gauge bosons, sleptons or top quarks. The decays of the superpartners can lead to many experimental signatures that may lead to different lepton, jet, and b -tagged jet multiplicities. To exploit this wide range of possible signatures, the analysis uses six R -parity-conserving SUSY scenarios featuring gluino, bottom squark or top squark pair production. These scenarios were used as benchmarks to identify regions of the phase space where the analysis can bring particularly useful complementarity to other SUSY searches, and sub-

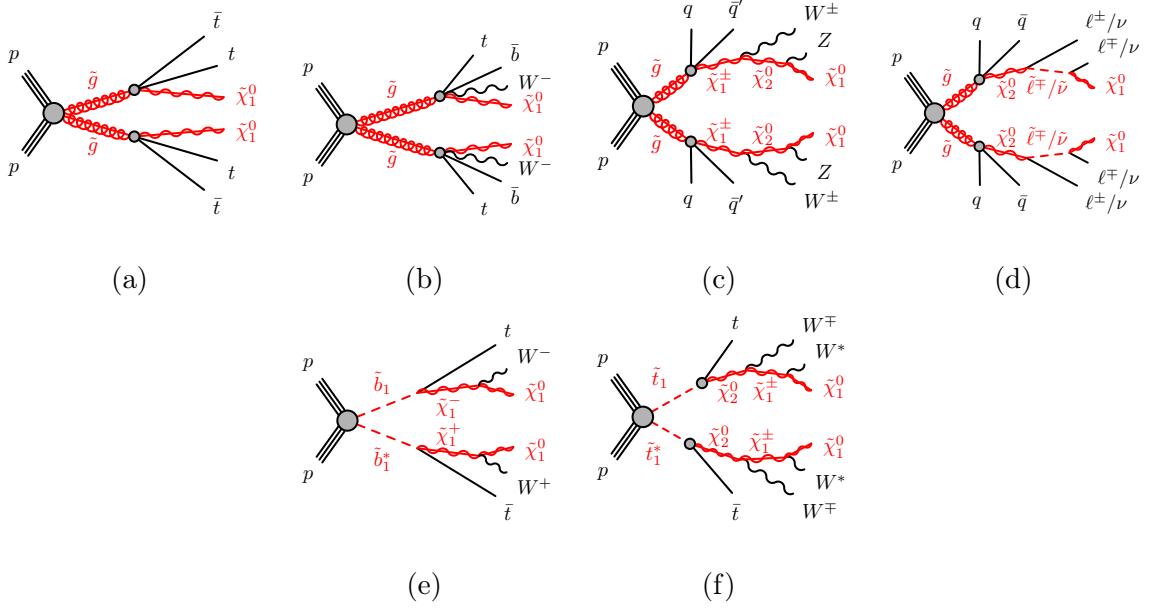


FIGURE 5.1: SUSY processes featuring gluino ((a), (b), (c), (d)) or third-generation squark ((e), (f)) pair production studied in this analysis. In Figure 5.1d, $\tilde{\ell} \equiv \tilde{e}, \tilde{\mu}, \tilde{\tau}$ and $\tilde{\nu} \equiv \tilde{\nu}_e, \tilde{\nu}_\mu, \tilde{\nu}_\tau$. In Figure 5.1f, the W^* labels indicate largely off-shell W bosons – the mass difference between $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_1^0$ is around 1 GeV.

sequently define signal regions with a particular focus on these regions. In this section, the scenarios considered are presented with details about the assumed superpartner masses and decay modes. Exclusion limits obtained prior to the work of the author will also be shown to highlight the improvement in reach that this analysis brings.

Gluino pair production with slepton-mediated two-step decay $\tilde{g} \rightarrow q\bar{q}\ell\bar{\ell}\tilde{\chi}_1^0$

This scenario (Fig. 5.1d) features gluino pair-production with two-step decays via neutralinos $\tilde{\chi}_2^0$ and sleptons, $\tilde{g} \rightarrow q\bar{q}'\tilde{\chi}_2^0 \rightarrow q\bar{q}'(\tilde{\ell}\ell/\tilde{\nu}\nu) \rightarrow q\bar{q}'(\ell\ell/\nu\nu)\tilde{\chi}_1^0$. The

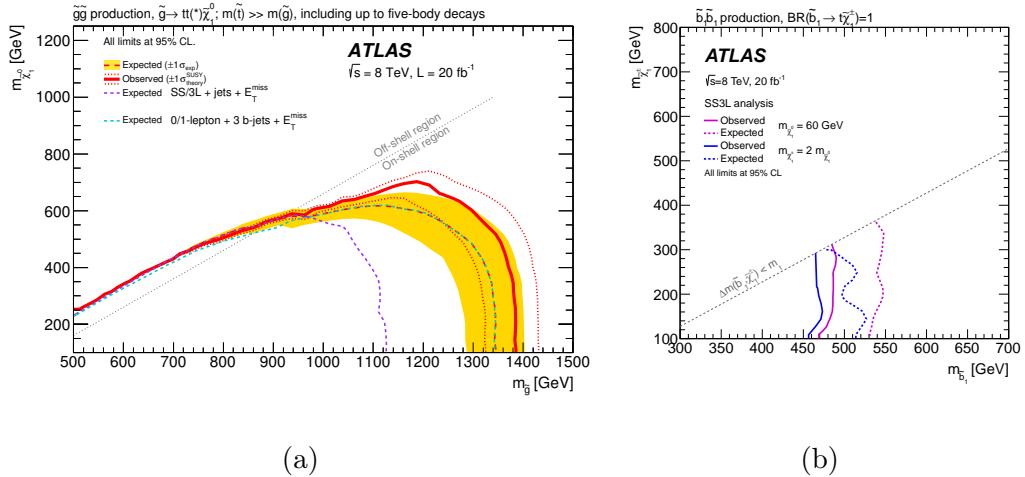


FIGURE 5.2: Exclusion limits on the gluino-stop offshell [36] (left) and direct sbottom [37] (right) scenarios set by ATLAS with the 2012 dataset prior to the author's work.

decays are mediated by generic heavy squarks, therefore the b -jet multiplicity in this scenario is low. The final state is made of charged leptons, four additional jets and invisible particles (neutrinos and neutralinos). The average jet multiplicity per event is the smallest among the four scenarios ; another characteristic is the large fraction of events with several leptons, unlike the other scenarios that have a rather low acceptance due to the branching ratios of $W \rightarrow \ell\nu$ or $Z \rightarrow \ell\ell$. The exclusion limits obtained in run-1 (Fig. 5.3b) show again that the SS/3L+jets final state is very competitive to probe those models. This scenario is used as a benchmark to define the signal regions with ≥ 3 leptons and no b -jet.

The signal grid is built with variable gluino and $\tilde{\chi}_1^0$ masses ; the $\tilde{\chi}_2^0$ mass is chosen half-way between the gluino and LSP masses, and the sleptons masses are also set equal and half-way between the $\tilde{\chi}_2^0$ and LSP masses. The $\tilde{\chi}_2^0$ may decay to

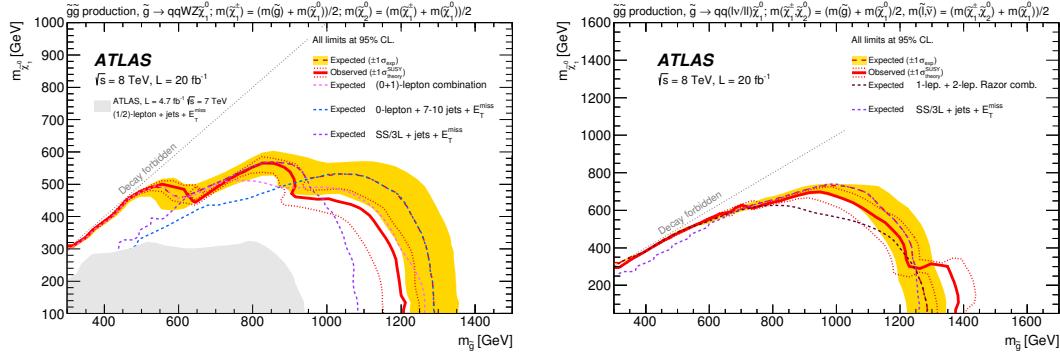


FIGURE 5.3: Exclusion limits on scenarios featuring gluino pair production followed by two-step decays via heavy gauge bosons or sleptons, set by ATLAS with the 2012 dataset prior to the author’s work [36].

any of the six “left-handed” sleptons (ℓ , $\tilde{\nu}$) with equal probability. “Right-handed” sleptons are assumed heavy and do not participate to the decay.

Gluino pair production with gaugino-mediated two-step decay $\tilde{g} \rightarrow$

$$q\bar{q}'WZ\tilde{\chi}_1^0$$

This scenario (Fig. 5.1c) features gluino pair-production with two-step decays via gauginos and W and Z bosons, $\tilde{g} \rightarrow q\bar{q}'\tilde{\chi}_1^\pm \rightarrow q\bar{q}'W\tilde{\chi}_2^0 \rightarrow q\bar{q}'WZ\tilde{\chi}_1^0$, mediated by generic heavy squarks of the first and second generations. The final state is made of two W and two Z bosons (possibly offshell), four additional jets and invisible particles (neutrinos and neutralinos). This generally leads to events with large jet multiplicities and a fair branching ratio for dileptonic final states. The exclusion limits obtained in run-1 indeed illustrate the competitiveness of the SS/3L+jets search (Fig. 5.3a) particularly the heavy- $\tilde{\chi}_1^0$ region of the phase space.

This scenario is used as a benchmark to define the signal regions with many jets but none tagged as a b -jet.

The signal grid is built with variable gluino and $\tilde{\chi}_1^0$ masses, and the $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_2^0$ masses are set such that the former lies half-way between the gluino and $\tilde{\chi}_1^0$ masses, and the latter half-way between $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_1^0$ masses.

Sbottom pair production with one-step decay $\tilde{b}_1 \rightarrow t\tilde{\chi}_1^\pm$

In this scenario (Fig. 5.1e), bottom squarks are rather light and assumed to decay in a top quark and a chargino $\tilde{\chi}_1^\pm$, with a subsequent $\tilde{\chi}_1^\pm \rightarrow W^\pm \tilde{\chi}_1^0$ decay, providing complementarity to the mainstream search [38] which focuses on the channel $\tilde{b}_1 \rightarrow b\tilde{\chi}_1^0$. The final state resulting from the production of a $\tilde{b}_1\tilde{b}_1^*$ pair contains two top quarks, two W bosons and two neutralinos. While this final state may lead to various experimental signatures, the model was considered in Run-1 [36] only by the same-sign leptons and jets search, leading to the exclusion limits presented in Fig. 5.2. Signal events typically contain one or two b -tagged jets, therefore this scenario is used as benchmark to define the signal regions with ≥ 1 b -jet.

The model adopts a fixed chargino-neutralino mass difference of 100 GeV, therefore always allowing on-shell W bosons in the $\tilde{\chi}_1^\pm \rightarrow W\tilde{\chi}_1^0$ decay¹. Only pair production of the lightest sbottom is considered, followed by an exclusive decay

1. A different chargino mass assumption is adopted in the current work compared to the Run-1 paper [36]. Fig. 5.2 is shown for illustration only. The reduced chargino-neutralino mass gap in the current analysis allows to study signal scenarios with heavy neutralinos, which were not considered previously.

in the aforementioned channel.

Gluino pair production with stop-mediated decay $\tilde{g} \rightarrow t\bar{t}\tilde{\chi}_1^0$

In this scenario inspired by naturalness arguments, gluinos are coupling preferentially to stops which are lighter than the other squarks. Gluinos are however considered lighter than stops, and decay directly into a $t\bar{t}\tilde{\chi}_1^0$ triplet via a virtual stop (Fig. 5.1a). The pair production of gluinos leads to a final state containing four top quarks and two neutralinos. This characteristic final state is accessible through various experimental signatures, which is why this model is commonly used as a benchmark to compare analyses sensitivities. The searches performed with Run-1 data [36], summarized in Fig. 5.2a, showed that the same-sign leptons final state is competitive only at large neutralino mass. This region of the phase space is consequently given a particular attention in the choice of signal regions described further on. For instance, the region of phase-space with $\Delta m(\tilde{g}, \tilde{\chi}_1^0) < 2m_t$, where gluinos decay via one or two offshell top quarks, is only accessible for this analysis. In the signal samples referenced in this document, the mass of the lightest stop is fixed to 10 TeV and is mostly a \tilde{t}_R state. Only gluino pair production is considered, followed by an exclusive decay in the aforementioned channel. Signal events typically contain many b -tagged jets, therefore this scenario is used as benchmark to define the signal regions with ≥ 2 b -jets.

$\tilde{t}_1\tilde{t}_1^*$ with “three-same-sign leptons” signature

Inspired by Ref. [39], a simplified model featuring a stop pair-production with two-step decays via a neutralino $\tilde{\chi}_2^0$ and a chargino $\tilde{\chi}_1^\pm$ is added in this version of the analysis, according to the decay illustrated on Fig. 5.1f :

$$\tilde{t}_1 \rightarrow t\tilde{\chi}_2^0 \rightarrow t\tilde{\chi}_1^\pm W^\mp \rightarrow tW^\pm W^\mp \tilde{\chi}_1^0.$$

This simplified model is a well-motivated representation of a pMSSM model. The lightest stop (\tilde{t}_1) is right-handed and $\tilde{\chi}_2^0$ is bino-like which leads to a large branching ratio in the decay $\tilde{t}_1 \rightarrow t\tilde{\chi}_2^0$. Furthermore, the decay $\tilde{\chi}_2^0 \rightarrow \tilde{\chi}_1^\pm W^\mp$ is also enhanced since $\tilde{\chi}_1^\pm$ is wino-like, as long as $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_1^0$ are nearly mass degenerate and $m_{\tilde{\chi}_2^0} - m_{\tilde{\chi}_1^0} < m_H = 125$ GeV to suppress the decay $\tilde{\chi}_2^0 \rightarrow \tilde{\chi}_1^0 + H$ (the decay $\tilde{\chi}_2^0 \rightarrow \tilde{\chi}_1^0 + Z$ is suppressed). By respecting these conditions and evading the bottom squark limit shown in Fig. 5.2b, we consider a one-dimensional grid with a \tilde{t}_1 mass varying between 550 GeV and 800 GeV with a 50 GeV gap², a two body decay to an on-shell top quark and a $\tilde{\chi}_2^0$ which has a 100 GeV mass difference from $\tilde{\chi}_1^0$. The mass difference between the $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_1^0$ is taken to be 500 MeV which is not excluded by the disappearing track analysis. In fact, this mass gap could easily be increased by introducing a small amount of higgsino mixing [40].

While the stop pair production is similar to the sbottom pair production in terms of kinematics, the stop pair production offers a unique topology that leads to three leptons of the same electric charge. This final state benefits from

2. Only the points at \tilde{t}_1 mass of 550 GeV are available at the moment.

an extreme reduction of the SM background while maintaining a good signal acceptance which helps loosen the kinematic cuts to access a more compressed SUSY phase space. As a result, this scenario is complementary to the search for bottom squarks.

Non-Universal Higgs Models

In references [41, 42, 43], theorists studied a complete two-extra-parameter non-universal Higgs model (NUHM2) that can have low fine tuning (natural) and predicts final state signatures that allow large background rejection while retaining high signal efficiency. The NUHM2 model allows the soft SUSY breaking masses of the Higgs multiplets, m_{H_u} and m_{H_d} , to be different from matter scalar masses (m_0) at the grand unification scale. The NUHM2 model is expected to form the effective theory for energies lower than m_{GUT} resulting from SU(5) or general SU(10) grand unified theories. The scalar mass m_0 , the soft SUSY breaking gaugino mass $m_{1/2}$, the pseudoscalar Higgs boson mass m_A , the trilinear SUSY breaking parameter A_0 , the weak scale ratio of Higgs field vacuum expectation values $\tan \beta$, and the superpotential Higgs mass μ are the free parameters. Both $m_{1/2}$ and μ are varied while the other parameters are fixed to $m_0 = 5$ TeV, $A_0 = -1.6m_0$, $\tan \beta = 15$, $m_A = 1$ TeV, and $\text{sign}(\mu) > 0$. These parameter choices lead directly to a Higgs mass of 125 GeV in accord with experiment. In this “radiatively-driven natural” SUSY approach, the higgsino is required to have mass below 200-300 GeV, the stop to have a mass below ~ 3 TeV, and the gluino below ~ 4 TeV. The model mainly involves gluino pair production with gluinos decaying predominantly to

Decay	BR	Decay	BR
$t\bar{t}\chi_1^0$	0.13	$t\bar{b}\chi_1^\pm$	0.45
$t\bar{t}\chi_2^0$	0.21	$t\bar{b}\chi_2^\pm$	0.04
$t\bar{t}\chi_3^0$	0.13	-	-
$t\bar{t}\chi_4^0$	0.02	-	-
$t\bar{t}\chi_i^0$	0.49	$t\bar{b}\chi_i^\pm$	0.49

TABLE 5.1: The dominant gluino decay modes for $m_{1/2} = 400$ GeV for the NUHM2 model.

$t\bar{t}\tilde{\chi}_1^0$ and $t\bar{b}\tilde{\chi}_1^\pm$, giving rise to final states with two same-sign leptons and E_T^{miss} . Table 5.1 shows the branching ratios of the dominant gluino decay modes for $m_{1/2} = 400$ GeV. Simulated NUHM2 signal samples with mass ($m_{1/2}$) values from 300-800 GeV and $\mu = 150$ GeV were generated where the gluino mass in this model is approximately $2.5 \times m_{1/2}$.

5.3 Dataset and Simulated Event Samples

5.3.1 Collision Data

The analysis uses pp -collisions data at $\sqrt{s} = 13$ TeV collected by the ATLAS detector during 2015 and 2016 with a peak instantaneous luminosity of $L = 1.4 \times 10^{34} \text{ cm}^{-2}\text{s}^{-1}$. The total integrated luminosity considered corresponds to 36.1 fb^{-1} (3.2 fb^{-1} in 2015 and 32.9 fb^{-1} in 2016) recorded after applying beam,

detector, and data-quality requirements. The combined luminosity uncertainty for 2015 and 2016 is 3.2%, assuming partially correlated uncertainties in 2015 and 2016. The integrated luminosity was established following the same methodology as that detailed in Ref. [44], from a preliminary calibration of the luminosity scale using a pair of x - y beam separation scans.

5.3.2 Simulated Event Samples

Monte Carlo (MC) simulated event samples are used to model the SUSY signal and SM backgrounds. The irreducible SM backgrounds refer to processes that lead to two same-sign and/or three “prompt” leptons where the prompt leptons are produced directly in the hard-scattering process, or in the subsequent decays of W, Z, H bosons or prompt τ leptons. The reducible backgrounds, mainly arising from $t\bar{t}$ and $V+jets$ production , are estimated either from data or from MC simulation as described in Section 6.

Table 5.2 presents the event generator, parton shower, cross-section normalization, PDF set and the set of tuned parameters for the modelling of the parton shower, hadronization and underlying event. Apart from the MC samples produced by the SHERPA generator, all MC samples used the EVTGEN v1.2.0 program [45] to model the properties of bottom and charm hadron decays.

The MC samples were processed through either a full ATLAS detector simulation [56] based on GEANT4 [24] or a fast simulation using a parameterization of the calorimeter response and GEANT4 for the ID and MS [25], and are reconstructed in the same manner as the data. All simulated samples are generated with a

Physics process	Event generator	Parton shower	Cross-section order	Cross-section value (fb)	PDF set	Set of tuned parameters
Signal	aMC@NLO 2.2.3 [22]	PYTHIA 8.186 [23]	NLO+NLL	See Table 5.4	NNPDF2.3LO [46]	A14 [47]
$t\bar{t} + X$						
$t\bar{t}W, t\bar{t}Z/\gamma^*$	aMC@NLO 2.2.2	PYTHIA 8.186	NLO [48]	600.8, 123.7	NNPDF2.3LO	A14
$t\bar{t}H$	aMC@NLO 2.3.2	PYTHIA 8.186	NLO [48]	507.1	NNPDF2.3LO	A14
$4t$	aMC@NLO 2.2.2	PYTHIA 8.186	NLO [22]	9.2	NNPDF2.3LO	A14
Diboson						
ZZ, WZ	SHERPA 2.2.1 [49]	SHERPA 2.2.1	NLO [50]	$1.3 \cdot 10^3, 4.5 \cdot 10^3$	NNPDF2.3LO	SHERPA default
inc. $W^\pm W^\pm$	SHERPA 2.1.1	SHERPA 2.1.1	NLO [50]	86	CT10 [51]	SHERPA default
Rare						
$t\bar{t}WW, t\bar{t}WZ$	aMC@NLO 2.2.2	PYTHIA 8.186	NLO [22]	9.9, 0.36	NNPDF2.3LO	A14
$tZ, tWZ, t\bar{t}t$	aMC@NLO 2.2.2	PYTHIA 8.186	LO	240, 16, 1.6	NNPDF2.3LO	A14
WH, ZH	aMC@NLO 2.2.2	PYTHIA 8.186	NLO [52]	$1.4 \cdot 10^3, 868$	NNPDF2.3LO	A14
Triboson	SHERPA 2.1.1	SHERPA 2.1.1	NLO [50]	14.9	CT10	SHERPA default
Irreducible (Incl.)						
$W+Jets$	POWHEG-BOX	PYTHIA 8.186	NNLO	$2.0 \cdot 10^7$	CT10	AZNLO[53]
$Z+Jets$	POWHEG-BOX	PYTHIA 8.186	NNLO	$1.9 \cdot 10^7$	CT10	AZNLO[53]
$t\bar{t}$	POWHEG-BOX	PYTHIA 6.428	NNLO+NNLL [54]	$8.3 \cdot 10^5$	CT10	PERUGIA2012 (P2012) [55]

TABLE 5.2: Simulated signal and background event samples : the corresponding event generator, parton shower, cross-section normalization, PDF set and set of tuned parameters are shown for each sample. Because of their very small contribution to the signal-region background estimate, $t\bar{t}WW, t\bar{t}WZ, tZ, tWZ, t\bar{t}t, WH, ZH$ and triboson are summed and labelled “rare” in the following.

range of minimum-bias interactions using PYTHIA 8 [23] with the MSTW2008LO PDF set [57] and the A2 tune overlaid on the hard-scattering event to account for the multiple pp interactions in the same bunch crossing (in-time pileup) and neighbouring bunch crossing (out-of-time pileup). The distribution of the average number of interactions per bunch crossing $\langle\mu\rangle$ ranges from 0.5 to 39.5, with a profile set as an estimate of the combined 2015+2016 data $\langle\mu\rangle$ profile. With larger luminosity collected during this year and the μ distribution in data being closer to that in the MC profile, the simulated samples are re-weighted to reproduce the

observed distribution of the average number of collisions per bunch crossing (μ).

Background process simulation

The two dominant irreducible background processes are $t\bar{t}V$ (with V being a W or Z/γ^* boson) and diboson production with final states of four charged leptons ℓ ,³ three charged leptons and one neutrino, or two same-sign charged leptons and two neutrinos.

The production of a $t\bar{t}V$ constitutes the main source of background with prompt same-sign leptons for event selections including b -jets. Simulated events for these processes were generated at NLO with AMC@NLO v2.2.2 [22] interfaced to PYTHIA 8, with up to two (ttW) or one ($ttZ^{(*)}$) extra parton included in the matrix elements [58]. The samples are normalised to the inclusive process NLO cross-section using appropriate k -factors [22].

The production of multiple W, Z bosons decaying leptonically constitutes the main source of background with prompt same-sign leptons for event selections vetoing b -jets. Diboson processes with four charged leptons, three charged leptons and one neutrino, or two charged leptons and two neutrinos were simulated at NLO using the SHERPA 2.2.1 generator [49], as described in detail in Ref. [50]. The main samples simulate $qq \rightarrow VV \rightarrow$ leptons production including the doubly resonant WZ and ZZ processes, non-resonant contributions as well as Higgs-mediated contributions, and their interferences ; up to three extra partons were

³. All lepton flavours are included here and τ leptons subsequently decay leptonically or hadronically.

included (at LO) in the matrix elements. Simulated events for the $W^\pm W^\pm jj$ process (including non-resonant contributions) were produced at LO with up to one extra parton, separately for QCD-induced ($\mathcal{O}(\alpha_{\text{em}}^4)$) and VBS-induced ($\mathcal{O}(\alpha_{\text{em}}^6)$) production – the interferences being neglected. Additional samples for VBS-induced $qq \rightarrow 3\ell\nu jj$ and $qq \rightarrow 4\ell$ and loop-induced $gg \rightarrow WZ^{(*)}/ZZ^{(*)}$ processes were also produced with the same configuration. The samples generated at NLO are directly normalized to the cross-sections provided by the generator.

Production of a Higgs boson in association with a $t\bar{t}$ pair is simulated using AMC@NLO [22] (in MADGRAPH v2.2.2) interfaced to HERWIG 2.7.1 [59]. The UEEE5 underlying-event tune is used together with the CTEQ6L1 [60] (matrix element) and CT10 [51] (parton shower) PDF sets. Simulated samples of SM Higgs boson production in association with a W or Z boson are produced with PYTHIA 8.186, using the A14 tune and the NNPDF23LO PDF set. Events are normalised with cross-sections calculated at NLO [52].

MADGRAPH v2.2.2 [61] is used to simulate the $t\bar{t}WW$, tZ , $t\bar{t}t\bar{t}$ and $t\bar{t}t$ processes, and the generator cross-section is used for tZ and $t\bar{t}t$. MADGRAPH interfaced to PYTHIA 8 is used to generate $t\bar{t}WZ$ processes, and appropriate k -factors are taken from [22]. AMC@NLO interfaced to PYTHIA 8 is used for the generation of the tWZ process, with an alternative sample generated with AMC@NLO interfaced to HERWIG used to evaluate the parton shower uncertainty. Fully leptonic triboson processes (WWW , WWZ , WZZ and ZZZ) with up to six charged leptons are simulated using SHERPA v2.1.1 and described in Ref. [50]. The 4ℓ and $2\ell + 2\nu$ processes are calculated at next-to-leading order (NLO) for up to one additional

parton ; final states with two and three additional partons are calculated at leading order (LO). The $WWZ \rightarrow 4\ell + 2\nu$ or $2\ell + 4\nu$ processes are calculated at LO with up to two additional partons. The $WWW/WZZ \rightarrow 3\ell + 3\nu$, $WZZ \rightarrow 5\ell + 1\nu$, $ZZZ \rightarrow 6\ell + 0\nu$, $4\ell + 2\nu$ or $2\ell + 4\nu$ processes are calculated is calculated at NLO with up to two extra partons at LO. The CT10 [51] parton distribution function (PDF) set is used for all SHERPA samples in conjunction with a dedicated tuning of the parton shower parameters developed by the SHERPA authors. The generator cross-sections (at NLO for most of the processes) are used when normalising these backgrounds.

Double parton scattering (DPS) occurs when two partons interact simultaneously in a proton-proton collision leading to two hard scattering processes overlapping in a detector event. Accordingly, two single W production processes can lead to a $W^\pm + W^\pm$ final state via DPS. This background is expected to have a negligible contribution to signal regions with high jet multiplicities. To estimate a conservative upper bound on cross-section for WW events which might arise from DPS, a standard ansatz is adopted : in this, for a collision in which a hard process (X) occurs, the probability that an additional (distinguishable) process (Y) occurs is parametrized as :

$$\sigma_{XY}^{DPS} = \sigma_X \sigma_Y / \sigma_{eff} \quad (5.1)$$

where σ_X is the production cross section of the hard process X and σ_{eff} (effective area parameter) parametrizes the double-parton interaction part of the production cross section for the composite system (X+Y). A value of σ_{eff} is 10-20 mb is

assumed in this study (as obtained from 7 TeV measurements, and with no observed dependence on \sqrt{s}), and it is independent on the processes involved. For the case of $W^\pm + W^\pm$ production :

$$\sigma_{W^\pm W^\pm}^{DPS} = \frac{\sigma_{W^+}\sigma_{W^+} + \sigma_{W^-}\sigma_{W^-} + 2\sigma_{W^+}\sigma_{W^-}}{\sigma_{eff}} \simeq 0.19 - 0.38 \text{ pb.} \quad (5.2)$$

After the application of the SR criteria, only 4 raw MC events in the DPS $WW \rightarrow \ell\nu\ell\nu$ remain. Table 5.3 shows the expected contribution in the SRs where some MC event survives all the cuts. The ranges quoted in the tables reflect the range in the predicted $\sigma_{W^\pm W^\pm}^{DPS}$ cross-section above, as well as the combinatorics for scaling the jet multiplicity⁴. Due to the large uncertainties involved in these estimates, some of them difficult to quantify (such as the modelling of DPS by PYTHIA at LO), the contribution from this background is not included in the final SR background estimates. Note that the estimated DPS contribution is typically much smaller than the uncertainty on the total background for the SRs.

Signal cross-sections and simulations

The signal processes are generated from leading order (LO) matrix elements with up to two extra partons (only one for the grid featuring slepton-mediated gluino decays), using the MADGRAPH v5.2.2.3 generator [22] interfaced to PYTHIA 8.186 [23] with the *ATLAS 14* tune [47] for the modelling of the SUSY decay chain, parton showering, hadronization and the description of the underlying

4. For instance, a DPS event with 6 jets can be due to the overlap of two events with 6+0 jets, or 5+1, 4+2 or 3+3 jets. All possible combinations are considered and the range quoted in the table shows the combinations leading to the smallest and largest correction factors.

TABLE 5.3: Number of raw MC events and its equivalent for 36.1 fb^{-1} with and without the correction as a function of the jet multiplicity. Only the SRs where at least one MC event passes all the cuts are shown.

SR	Raw MC events	Without N_{jet} correction	With N_{jet} correction
Rpc2L0bS	2	0.016-0.033	0.09-0.38
Rpc2L0bH	1	0.006-0.012	0.05-0.17

TABLE 5.4: Signal cross-sections [pb] and related uncertainties [%] for scenarios featuring $\tilde{g}\tilde{g}$ (top table) or $\tilde{b}_1\tilde{b}_1^*$ (bottom table) production, as a function of the pair-produced superpartner mass, reproduced from Ref. [62].

Gluino mass (GeV)	500	550	600	650	700
Cross section (pb)	$27.4 \pm 14\%$	$15.6 \pm 14\%$	$9.20 \pm 14\%$	$5.60 \pm 14\%$	$3.53 \pm 14\%$
750	800	850	900	950	1000
$2.27 \pm 14\%$	$1.49 \pm 15\%$	$0.996 \pm 15\%$	$0.677 \pm 16\%$	$0.466 \pm 16\%$	$0.325 \pm 17\%$
1050	1100	1150	1200	1250	1300
$0.229 \pm 17\%$	$0.163 \pm 18\%$	$0.118 \pm 18\%$	$0.0856 \pm 18\%$	$0.0627 \pm 19\%$	$0.0461 \pm 20\%$
1350	1400	1450	1500	1550	1600
$0.0340 \pm 20\%$	$0.0253 \pm 21\%$	$0.0189 \pm 22\%$	$0.0142 \pm 23\%$	$0.0107 \pm 23\%$	$0.00810 \pm 24\%$

Sbottom mass (GeV)	400	450	500	550
Cross section (pb)	$1.84 \pm 14\%$	$0.948 \pm 13\%$	$0.518 \pm 13\%$	$0.296 \pm 13\%$
600	650	700	750	800
$0.175 \pm 13\%$	$0.107 \pm 13\%$	$0.0670 \pm 13\%$	$0.0431 \pm 14\%$	$0.0283 \pm 14\%$

event. Parton luminosities are provided by the NNPDF23LO [63] set of parton distribution functions. Jet-parton matching is realized following the CKKW-L

prescription [64], with a matching scale set to one quarter of the pair-produced superpartner mass.

The signal samples are normalised to the next-to-next-to-leading order cross-section from Ref. [62] including the re-summation of soft gluon emission at next-to-next-to-leading-logarithmic accuracy (NLO+NLL), as detailed in Ref. [65]; some of these cross-sections are shown for illustration in Table 5.4.

Cross-section uncertainties are also taken from Ref. [62] as well, and include contributions from varied normalization and factorization scales, as well as PDF uncertainties. They typically vary between 15 and 25%. Uncertainties on the signal acceptance are not considered since these are generally smaller than the uncertainties on the inclusive production cross-section.

5.4 Event Selection

5.4.1 Pre-selection and event cleaning

A sample of two same-sign or three leptons is selected applying the following criteria :

- **Jet Cleaning** : Events are required to pass a set of cleaning requirements.

An event is rejected if any pre-selected jets ($|\eta| < 4.9$, after jet-electron overlap removal) fails the jet quality criteria. The cleaning requirements are intended to remove events where significant energy was deposited in the calorimeters due to instrumental effects such as cosmic rays, beam-induced (non-collision) particles, and noise. Around 0.5% of data events are lost

after applying this cut.

- **Primary Vertex** : Events are required to have a reconstructed vertex [66] with at least two associated tracks with $p_T > 400$ MeV. The vertex with the largest Σp_T^2 of the associated tracks is chosen as the primary vertex of the event. This cut is found to be 100% efficient.
- **Bad Muon Veto** : Events containing at least one pre-selected muon satisfying $\sigma(q/p)/|q/p| > 0.2$ before the overlap removal are rejected. Around 0.1% of data events are removed by this cut.
- **Cosmic Muon Veto** : Events containing a cosmic muon candidate are rejected. Cosmic muon candidates are looked for among pre-selected muons, if they fail the requirements $|z_0| < 1.0$ mm and $|d_0| < 0.2$ mm, where the longitudinal and transverse impact parameters z_0 and d_0 are calculated with respect to the primary vertex. Up to 6% of data events are lost at this cleaning cut.
- **At least two leptons** : Events are required to contain at least two signal leptons with $p_T > 20$ GeV for the two leading leptons. If the event contains a third signal lepton with $p_T > 10$ GeV the event is regarded as a three-lepton event, otherwise as a two-lepton event. The data sample obtained is then divided into three channels depending on the flavor of the two leptons forming a same-sign pair (ee , $\mu\mu$, $e\mu$). If more than one same-sign pairs can be built, the one involving the leading lepton will be considered for the channel selection.
- **Same-sign** : if the event has exactly two leptons, then these two leptons

have to be of identical electric charge (“same-sign”).

The following event variables are also used in the definition of the signal and validation regions in the analysis :

- The inclusive effective mass m_{eff} defined as the scalar sum of all the signal leptons p_T , all signal jets p_T and E_T^{miss} .

5.4.2 Trigger strategy

Events are selected using a combination of dilepton and E_T^{miss} triggers, the latter being used only for events with $E_T^{\text{miss}} > 250\text{GeV}$. Since the trigger thresholds have been raised between 2015 and 2016 due to the continuous increase of the instantaneous luminosity, the dilepton triggers used for :

- 2015 data : logical **or** of a trigger with two electrons of 12 GeV, with an electron of 17 GeV and a muon of 14 GeV, with two muons of 18 GeV and 8 GeV.
- 2016 data : logical **or** of a trigger with two electrons of 17 GeV, with an electron of 17 GeV and a muon of 14 GeV, with two muons of 22 GeV and 8 GeV.

The E_T^{miss} trigger was also raised from 70GeV to a 100 GeV and 110 GeV . The trigger-level requirements on E_T^{miss} and the leading and subleading lepton p_T are looser than those applied offline to ensure that trigger efficiencies are constant in the relevant phase space.

Trigger matching

For events exclusively selected via one or several of the dilepton triggers, we

require a matching between the online and offline leptons with $p_T > 20\text{GeV}$. with the exception of the di-muon trigger for which muons with $p_T > 10\text{GeV}$ are also considered. In addition, for the di-muon trigger in the 2016 configuration, the p_T requirement of the leading matched muon is raised to 23 GeVto remain on the trigger efficiency plateau.

Trigger scale factors

The simulated events are corrected for any potential differences in the trigger efficiency between data and MC simulation. Assuming uncorrelation between the E_T^{miss} and dilepton triggers, trigger scale factors are applied to MC events which were not selected by the E_T^{miss} trigger. These scale factors are computed for each event, considering the combination of fired triggers, the number and flavours of the leptons,

5.4.3 Object definition

This section presents the definitions of the objects used in the analysis : jets, electrons, muons and E_T^{miss} (the taus are not considered).

Jets

The jets are kept only if they have $p_T > 20\text{ GeV}$ and lie within $|\eta| < 2.8$. To mitigate the effects of pileup, the pile-up contribution is subtracted from the expected average energy contribution according to the jet area [67, 68]. In order to reduce the effects of pile-up, a significant fraction of the tracks in jets with $p_T < 60\text{GeV}$ and $|\eta| < 2.4$ must originate from the primary vertex, as defined by

the jet vertex tagger (JVT) [69]. The jet calibration follows the prescription in Ref. [68].

The 70% efficiency operating point of the MV2c10 algorithm was chosen which corresponds to the average efficiency for tagging b -jets in simulated $t\bar{t}$ events. This efficiency working point was favored by optimisation studies performed in simulated signal and background samples. The rejection factors for light-quark/gluon jets, c -quark jets and hadronically decaying τ leptons in simulated $t\bar{t}$ events are approximately 380, 12 and 54, respectively [71, 72]. Jets with $|\eta| < 2.5$ which satisfy the b -tagging and JVT requirements are identified as b -jets. Correction factors and uncertainties determined from data for the b -tagging efficiencies and mis-tag rates are applied to the simulated samples [71].

For the data-driven background estimations, two categories of electrons and muons are used : “candidate” and “signal” with the latter being a subset of the “candidate” leptons satisfying tighter selection criteria.

Electrons

Electron candidates are reconstructed from energy depositions in the electromagnetic calorimeter and required to be matched to an inner detector track, to have $p_T > 10\text{GeV}$ and $|\eta| < 2.47$, and to pass the “Loose” likelihood-based electron identification requirement [26]. Electrons in the transition region between the barrel and endcap electromagnetic calorimeters ($1.37 < |\eta| < 1.52$) are rejected to reduce the contribution from fake/non-prompt electrons. The transverse impact parameter d_0 with respect to the reconstructed primary vertex must satisfy

$|d_0/\sigma(d_0)| < 5$. This last requirement helps reduce the contribution from charge mis-identification.

Signal electrons are additionally required to pass the “Medium” likelihood-based identification requirement [26]. Only signal electrons with $|\eta| < 2.0$ are considered, to reduce the level of charge-flip background. In addition, signal electrons that are likely to be reconstructed with an incorrect charge assignment are rejected using a few electron cluster and track properties : the track impact parameter, the track curvature significance, the cluster width and the quality of the matching between the cluster and its associated track, both in terms of energy and position. These variables, as well as the electron p_T and η , are combined into a single classifier using a boosted decision tree (BDT). A selection requirement on the BDT output is chosen such as to achieve a rejection factor between 7 and 8 for electrons with a wrong charge assignment while selecting properly measured electrons with an efficiency of 97% (in $Z \rightarrow ee$ MC).

A multiplicative event weight is applied for each signal electron in MC to the overall event weight in order to correct for differences in efficiency between data and MC.

Muons

Muons candidates are reconstructed from muon spectrometer tracks matched to the inner detector tracks in the region $|\eta| < 2.5$. Muon candidates must pass the “Medium” identification requirements [73] and have $p_T > 10\text{GeV}$ and $|\eta| < 2.4$. Signal muons are required to pass $|d_0/\sigma(d_0)| < 3$ and $|z_0 \cdot \sin(\theta)| < 0.5\text{mm}$.

A multiplicative event weight is applied for each selected muon in MC to the overall event weight in order to correct for differences in efficiency between data and MC.

Overlap removal

According to the above definitions, one single final state object may fall in more than one category, being therefore effectively double-counted. For example, one isolated electron is typically reconstructed both as an electron and as a jet. A procedure to remove overlaps between final state objects was therefore put in place, and applied on pre-selected objects. Any jet within a distance $\Delta R_y \equiv \sqrt{(\Delta y)^2 + (\Delta\phi)^2} = 0.2$ of a lepton candidate is discarded, unless the jet is *b*-tagged,⁵ in which case the lepton is discarded since it probably originated from a semileptonic *b*-hadron decay. Any remaining lepton within $\Delta R_y \equiv \min\{0.4, 0.1 + 9.6\text{GeV}/p_T(\ell)\}$ of a jet is discarded. In the case of muons, the muon is retained and the jet is discarded if the jet has fewer than three associated tracks. This reduces inefficiencies for high-energy muons undergoing significant energy loss in the calorimeter.

Missing transverse energy

The missing transverse energy (E_T^{miss}) is computed as a negative vector sum of the transverse momenta of all identified candidate objects (electrons, photons [74], muons and jets) and an additional soft term. The soft term is constructed from

⁵. In this case the *b*-tagging operating point corresponding to an efficiency of 85% is used.

all tracks associated with the primary vertex but not with any physics object. In this way, the E_T^{miss} is adjusted for the best calibration of the jets and the other identified physics objects listed above, while maintaining approximate pile-up independence in the soft term [75, 76].

5.4.4 Data-MC comparisons

In order to validate the various choices made regarding the object definitions and event selection, check their sensible behavior and their reasonable modelling in the simulations, we looked at the distributions of several kinematic variables obtained with the full available data set. Figures 5.4-5.6 show such selected distributions in data compared to MC. The background distributions are taken directly from MC with no data-driven estimation of the charge flip or non-prompt lepton backgrounds.

Figure 5.4 shows the dilepton invariant mass distributions for both opposite-sign (OS) and same-sign (SS) dilepton events, computed with the two leading p_T leptons. A very good agreement with MC is observed in the OS channels, with a clear Z -boson mass peak in the ee and $\mu\mu$ channels. In the SS channels, the Z -boson mass peak is also observed in the ee channel due to electron charge mis-identification, with MC overestimating data. An accumulation of events at the Z -boson mass is also observed in the SS $e\mu$ and $\mu\mu$ channels due to three-lepton events from either $Z+jets$ with a fake lepton or from WZ production.

The transverse momentum distributions of the signal leptons used in the analysis are shown in Figure 5.5, with a reasonable data-MC agreement except

at low lepton p_T where some discrepancies and accumulation of events involving fake leptons ($Z + \text{jets}$, $W + \text{jets}$, $t\bar{t}$) are observed. Figure 5.6 shows the E_T^{miss} and m_{eff} distributions.

The background estimation, in Chapter 6 and Chapter 7, will be dedicated to improving the estimates of the background prediction using data-driven methods.

5.5 Signal Regions

In order to maximize the sensitivity to the signal models of Figure 5.1, 13 non-exclusive signal regions are defined in Table 5.6. The SRs are named in the form $RPCNLMbX$, where N indicates the number of leptons required, M the number of b -jets required, and X indicates the severity of the E_T^{miss} or m_{eff} requirements (Soft, Medium or Hard). All signal regions allow any number of additional leptons in addition to a $e^\pm e^\pm$, $e^\pm \mu^\pm$ or $\mu^\pm \mu^\pm$ pair. Signal regions with 3 leptons can be either any charge combination or all three with the same charge (Rpc3LSS1b). For each lepton/ b -jet multiplicity, two signal regions are defined targeting either compressed spectra or large mass splittings.

The optimization of the definitions of signal regions relied on a brute-force scan of several discriminating variables in a loose classification of events in terms of number of b -jets and/or leptons in the final state, each being associated to the signal scenario favouring this final state. The other main discriminant variables (e.g number of jets above a certain p_T threshold, m_{eff} , E_T^{miss} , $E_T^{\text{miss}}/m_{\text{eff}}$ ratio) were then allowed to vary, to determine for each point of the parameter space the

best configuration. The figure of merit used to rank configurations is the discovery significance (Z_0) defined in Eq. 5.3 which represents a statistical test based on a ratio of two Poisson means [77] :

$$Z_0 = \sqrt{2 \left((s + b) \ln \left(1 + \frac{s}{b} \right) - s \right)} \quad (5.3)$$

where s and b represent the expected number of signal and background events⁶. A realistic systematic uncertainty of $\Delta b = 30\%$ on the expected background yield was included in the statistical test by replacing b with $b + \Delta b$ in Eq. 5.3. To preserve the discovery potential, only configurations leading to at least two signal events were considered for a given signal point. The total number of background events should not be smaller than 1 ; to model in a more realistic way the effect of non-prompt and fake leptons and electron charge mis-identification backgrounds, which are determined from data in the analysis, the MC predictions for those processes in $t\bar{t}$ and $Z+jets$ MC were scaled using the factors obtained from the MC template method (Section 6.3.1), as shown in Table 5.5. Note that different corrections are applied depending on the showering (Pythia or Sherpa) used for each sample, and for the fake and non-prompt leptons originated from heavy-flavour (HF) and light-flavour (LF).

Since the signal regions defined out of the scanning procedure may not be mutually exclusive, the results expressed in terms of exclusion limits will be obtained by using for each signal point the SR giving the best expected sensitivity. For the latter, only the signal regions that were defined aiming for that particular

6. Note that Eq.5.3 simplifies to the commonly used figure of merit $\frac{s}{\sqrt{b}} + \mathcal{O}\left(\frac{s}{b}\right)$ if $\frac{s}{b} \ll 1$.

TABLE 5.5: Scaling factors applied to the electron charge-flip and non-prompt/fake lepton background in the SR optimization procedure.

	Charge mis-id	HF e	HF μ	LF e	LF μ
Pythia	0.96 ± 0.08	1.80 ± 0.45	2.10 ± 0.58	1.55 ± 0.14	0.74 ± 0.81
Sherpa	1.02 ± 0.09	2.72 ± 0.57	1.81 ± 0.75	1.16 ± 0.18	1.84 ± 1.16

signal model are considered, though.

Figures 5.7 and 5.8 show the performance of the SRs in the four RPC benchmark models with top quarks considered. The discovery significance for each signal point is shown, together with the contours corresponding to a 3σ discovery sensitivity, 1.64σ discovery sensitivity and 95% confidence level limits.

Dedicated new SRs have been optimized for the gluino pair production with stop-mediated decay $\tilde{g} \rightarrow t\bar{t}\tilde{\chi}_1^0$ with off-shell tops.

The $\tilde{g}\tilde{g}$ production with $\tilde{g} \rightarrow t\bar{t}\tilde{\chi}_1^0$ scenario at low LSP masses (where the multi- b analysis has a much better sensitivity [78]) is not the only motivation for Rpc2L2bH signal region, but also the NUHM2 model, which features large branching ratios for the $\tilde{g} \rightarrow t\bar{t}\tilde{\chi}_{1,2,3}^0$ and $\tilde{g} \rightarrow t\bar{b}\tilde{\chi}_{1,2}^\pm$ decays. As shown in Figure 5.9, with the Rpc2L2bH signal region $m_{1/2}$ values of 600 GeV can be excluded at 95% CL or observed with a 3σ significance. This SR will be then used for the first interpretation in this model.

In addition, the SS/3L analysis has the unique potential to explore the region of phase space at high LSP masses with a more compressed spectra. This scenario

Signal region	$N_{\text{leptons}}^{\text{signal}}$	$N_{b\text{-jets}}$	N_{jets}	p_T^{jet} [GeV]	E_T^{miss} [GeV]	m_{eff} [GeV]	$E_T^{\text{miss}}/m_{\text{eff}}$	Other	Targeted Signal
Rpc2L2bS	$\geq 2\text{SS}$	≥ 2	≥ 6	> 25	> 200	> 600	> 0.25	–	Fig. 5.1a
Rpc2L2bH	$\geq 2\text{SS}$	≥ 2	≥ 6	> 25	–	> 1800	> 0.15	–	Fig. 5.1a, NUHM2
Rpc2Lsoft1b	$\geq 2\text{SS}$	≥ 1	≥ 6	> 25	> 100	–	> 0.3	$20, 10 < p_T^{\ell_1}, p_T^{\ell_2} < 100 \text{ GeV}$	Fig. 5.1b
Rpc2Lsoft2b	$\geq 2\text{SS}$	≥ 2	≥ 6	> 25	> 200	> 600	> 0.25	$20, 10 < p_T^{\ell_1}, p_T^{\ell_2} < 100 \text{ GeV}$	Fig. 5.1b
Rpc2L0bS	$\geq 2\text{SS}$	$= 0$	≥ 6	> 25	> 150	–	> 0.25	–	Fig. 5.1c
Rpc2L0bH	$\geq 2\text{SS}$	$= 0$	≥ 6	> 40	> 250	> 900	–	–	Fig. 5.1c
Rpc3L0bS	≥ 3	$= 0$	≥ 4	> 40	> 200	> 600	–	–	Fig. 5.1d
Rpc3L0bH	≥ 3	$= 0$	≥ 4	> 40	> 200	> 1600	–	–	Fig. 5.1d
Rpc3L1bS	≥ 3	≥ 1	≥ 4	> 40	> 200	> 600	–	–	Other
Rpc3L1bH	≥ 3	≥ 1	≥ 4	> 40	> 200	> 1600	–	–	Other
Rpc2L1bS	$\geq 2\text{SS}$	≥ 1	≥ 6	> 25	> 150	> 600	> 0.25	–	Fig. 5.1e
Rpc2L1bH	$\geq 2\text{SS}$	≥ 1	≥ 6	> 25	> 250	–	> 0.2	–	Fig. 5.1e
Rpc3LSS1b	$\geq \ell^{\pm}\ell^{\pm}\ell^{\pm}$	≥ 1	–	–	–	–	–	veto $81 < m_{e^{\pm}e^{\pm}} < 101 \text{ GeV}$	Fig. 5.1f

TABLE 5.6: Summary of the signal region definitions. Unless explicitly stated in the table, at least two signal leptons with $p_T > 20 \text{ GeV}$ and same charge (SS) are required in each signal region. Requirements are placed on the number of signal leptons ($N_{\text{leptons}}^{\text{signal}}$), the number of b -jets with $p_T > 20 \text{ GeV}$ ($N_{b\text{-jets}}$), the number of jets (N_{jets}) above a certain p_T threshold (p_T^{jet}), E_T^{miss} , m_{eff} and/or $E_T^{\text{miss}}/m_{\text{eff}}$. The last column indicates the targeted signal model. The Rpc3L1b and Rpc3L1bH SRs are not motivated by a particular signal model and can be seen as a natural extension of the Rpc3L0b SRs with the same kinematic selections but requiring at least one b -jet.

leads to softer decay products, in particular softer b -jets as seen in Figure 5.10, which makes the multi- b analysis less sensitive. For this reason, two additional

signal regions were introduced with at least 1 b -jet (Rpc2Lsoft1b) or 2 b -jets (Rpc2Lsoft2b) defined in Table 5.6. In addition, these signal regions are defined with an upper cut on the leading lepton p_T . The sensitivity is degraded if this upper cut is removed as shown in Figure 5.11.

Motivated by the \tilde{t} production with $\tilde{t}_1 \rightarrow \tilde{\chi}_2^0 W$ model in Section 5.2, the signature of three leptons with the same electric charge (3LSS) is explored for the first time. As shown in Figure 5.12, after an inclusive 3LSS selection, the background is dominated by dibosons and $Z+jets$ (with only one real lepton, and the two other leptons with either an electron with charge mis-identified or a fake lepton) both dominantly without b -jets. Once a b -jet requirement is applied, the background is dominated by $t\bar{t}V$, with a clear peak at $m_{\ell\ell} \approx m_Z$ showing that a large fraction of these events are originated from charge mis-identification from events containing $Z \rightarrow ee$. After applying a $81 < m_{e^\pm e^\pm} < 101$ GeV veto, the background is reduced to only 1.7 events for 36.5 fb^{-1} , almost removing the $Z+jets$ and diboson backgrounds completely. The final background is dominated by $t\bar{t}+H, Z, W$, with $\approx 60\%$ originating from charge flips and $\approx 40\%$ from fakes and non-prompt leptons. With these very generic selections (Rpc3LSS1b in Table 5.6), a significance of 3.7σ can be obtained for $m_{\tilde{t}} = 550$ GeV. Figure 5.13 shows some lepton distributions, including the number of electrons, where most of the charge flip background populates the bins with 2 or 3 electrons, although cutting away those bins would also have a large impact on the signal.

Finally, since the SRs defined for the $\tilde{g}\tilde{g}$ production with $\tilde{g} \rightarrow q\bar{q}\ell\bar{\ell}\tilde{\chi}_1^0$ feature a b -jet veto (Rpc3L0bS and Rpc3L0bH), and to avoid leaving uncovered the 3

lepton plus b -jets signature, SRs with the same kinematic cuts as Rpc3L0bS and Rpc3L0bH but with a ≥ 1 b -jet requirement are also proposed in Table 5.6 as Rpc3L1bS and Rpc3L1bH.

5.6 Analysis Acceptance and Efficiency

Based on the signal regions defined in Section 5.6, it is useful to evaluate the signal acceptance of the analysis using parton level MC simulation (truth study). The acceptance encodes all the kinematic cuts applied on the signal as well as the branching ratios of all the decay particles. This information will help us understand the sensitivity reach of the analysis and also allow theorists to use this information when comparing the simplified model results to their models. The signal acceptance is shown in Figure 5.14a with the rest of the signal regions shown in Appendix A.4. Table 5.7 shows an example of a detailed cut-flow for weighted signal MC events illustrating the impact of cuts from the Rpc2L0bH signal region on a signal model. More tables are shown in Appendix A.3.

Another quantity of interest, to experimentalist in particular, is the detector efficiency that entails the reconstruction and identification efficiencies of the different particles used in the analysis. The efficiency ϵ can be obtained from the relation

$$S = L_{int} \cdot \sigma_{prod} \cdot A \cdot \epsilon \quad (5.4)$$

Where S is the expected number of signal events, σ_{prod} is the production cross section of the signal process, L_{int} is the integrated luminosity. Figure 5.14b shows

Rpc2L0bH, $\tilde{g}\tilde{g}$ production, $\tilde{g} \rightarrow q\bar{q}' WZ\tilde{\chi}_1^0$ $m_{\tilde{g}} = 1.6\text{TeV}$, $(m_{\tilde{\chi}_1^\pm} - 750) = (m_{\tilde{\chi}_2^0} - 375) = m_{\tilde{\chi}_1^0} = 100\text{GeV}$	
MC events generated	20000
Expected for 36.1 fb^{-1}	2.9×10^2
≥ 2 SS leptons ($p_T > 20\text{GeV}$)	12.8 ± 0.5
Trigger	12.5 ± 0.5
no b -jet ($p_T > 20\text{GeV}$)	8.5 ± 0.4
≥ 6 jets ($p_T > 40\text{GeV}$)	7.12 ± 0.35
$E_T^{\text{miss}} > 250\text{GeV}$	5.13 ± 0.29
$m_{\text{eff}} > 0.9\text{TeV}$	5.13 ± 0.29

TABLE 5.7: Number of signal events at different stages of the Rpc2L0bH signal region selection. Only statistical uncertainties are shown.

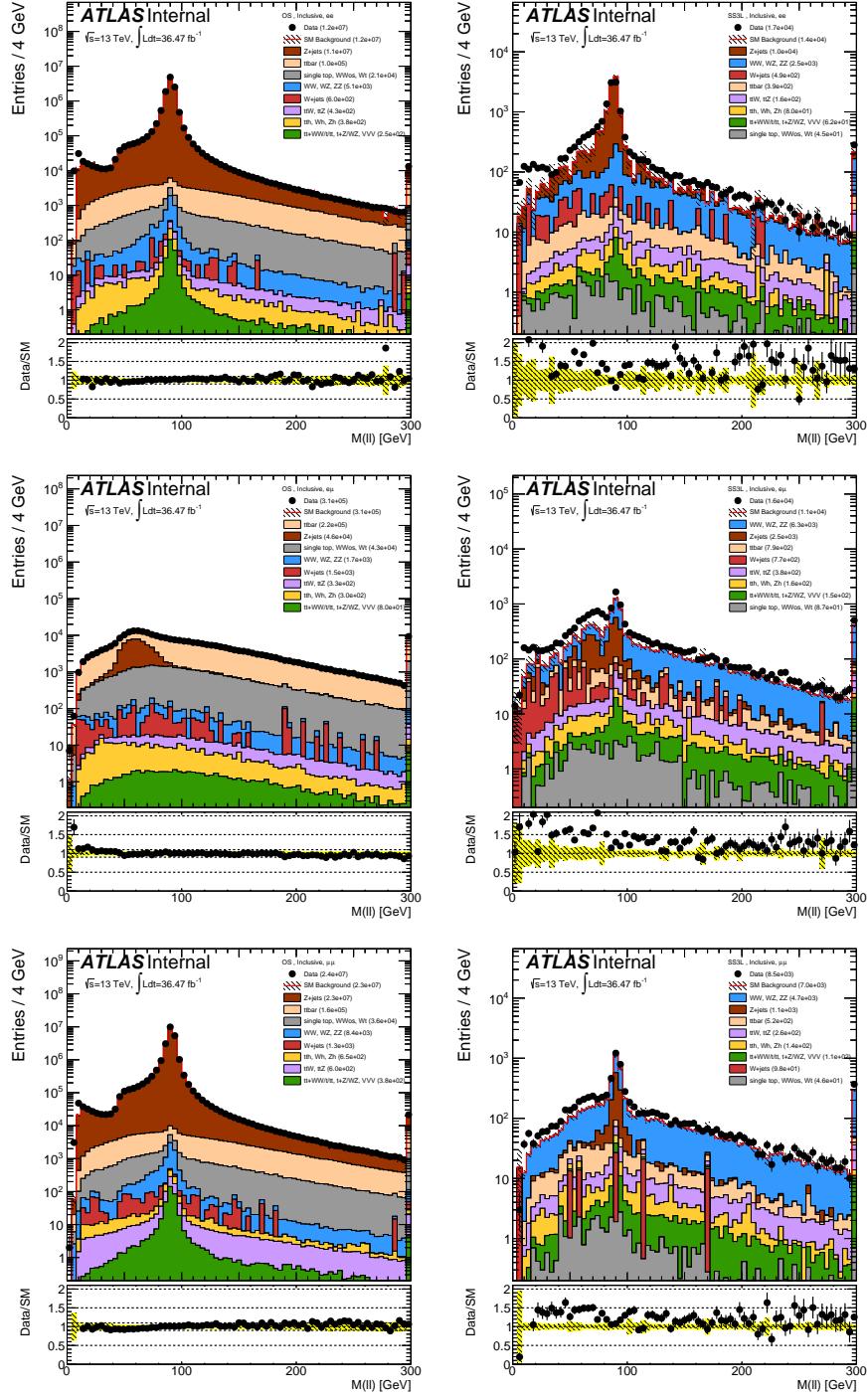


FIGURE 5.4: Dilepton invariant mass distributions for opposite-sign (left) and same-sign (right) pairs for events selected in the ee (top), $e\mu$ (center) and $\mu\mu$ (bottom) channels. No low-mass Drell-Yan sample is included. Only luminosity and MC statistical uncertainties are included.

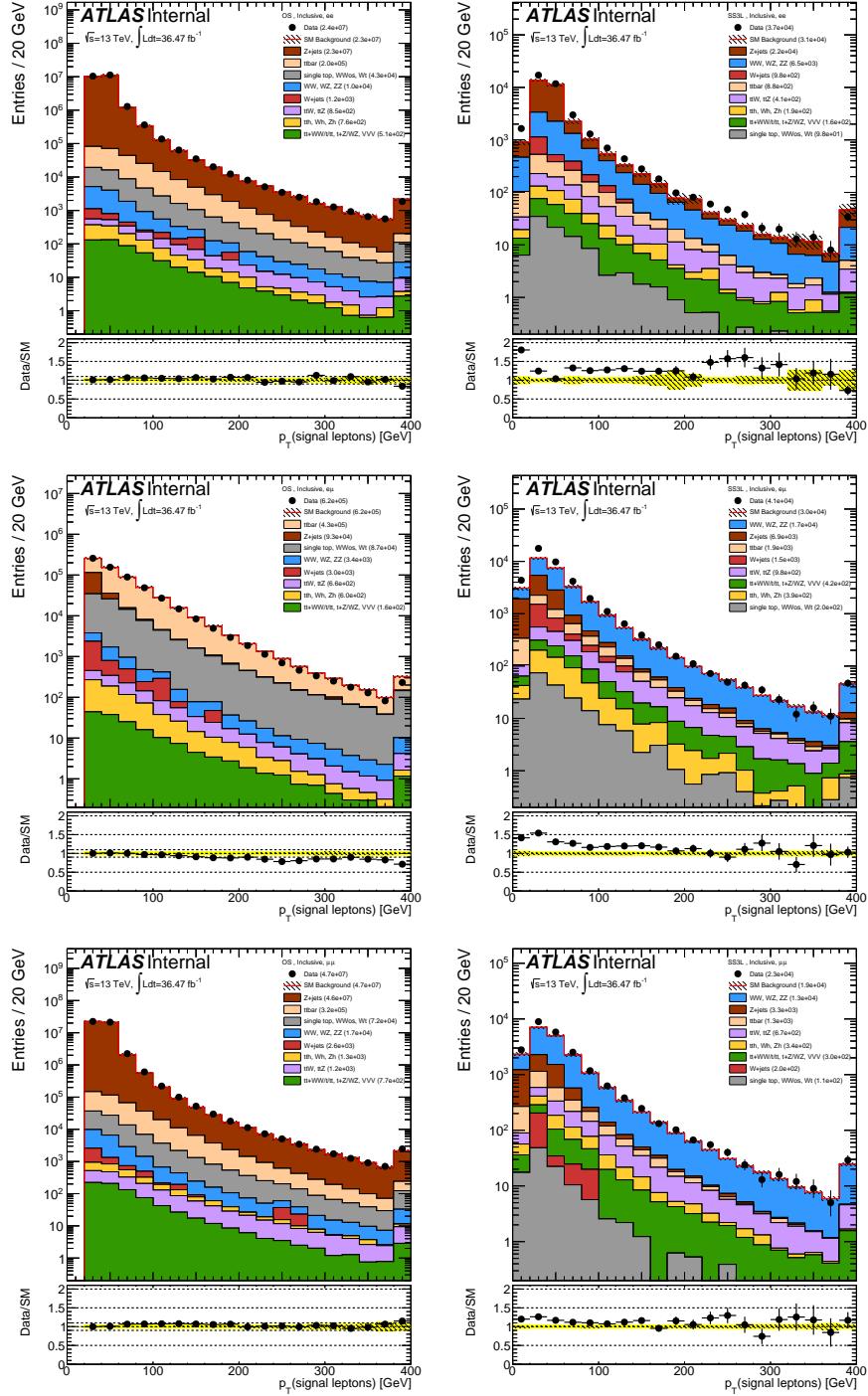


FIGURE 5.5: Signal lepton transverse momentum distributions for opposite-sign (left) and same-sign (right) pairs for events selected in the ee (top), $e\mu$ (center) and $\mu\mu$ (bottom) channels. The background contribution is taken directly from MC with no data-driven estimation of the background with fake and non-prompt leptons or charge mis-identification. Only luminosity and MC statistical uncertainties are included.

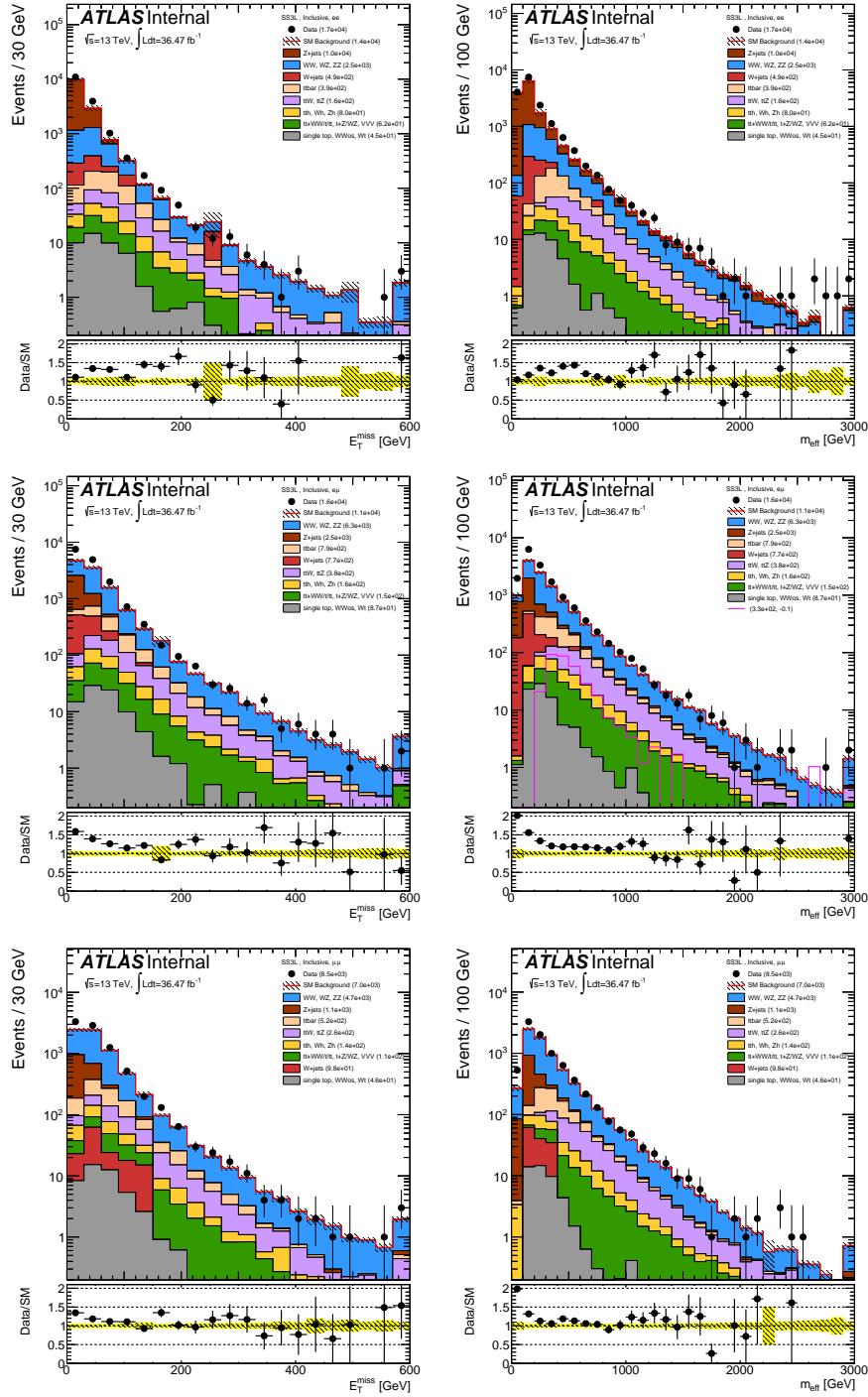


FIGURE 5.6: Distributions of the E_T^{miss} (left) and effective mass (right) for events selected in the ee (top), $e\mu$ (center) and $\mu\mu$ (bottom) channels. The background contribution is taken directly from MC with no data-driven estimation of the background with fake and non-prompt leptons or charge mis-identification. Only luminosity and MC statistical uncertainties are included.

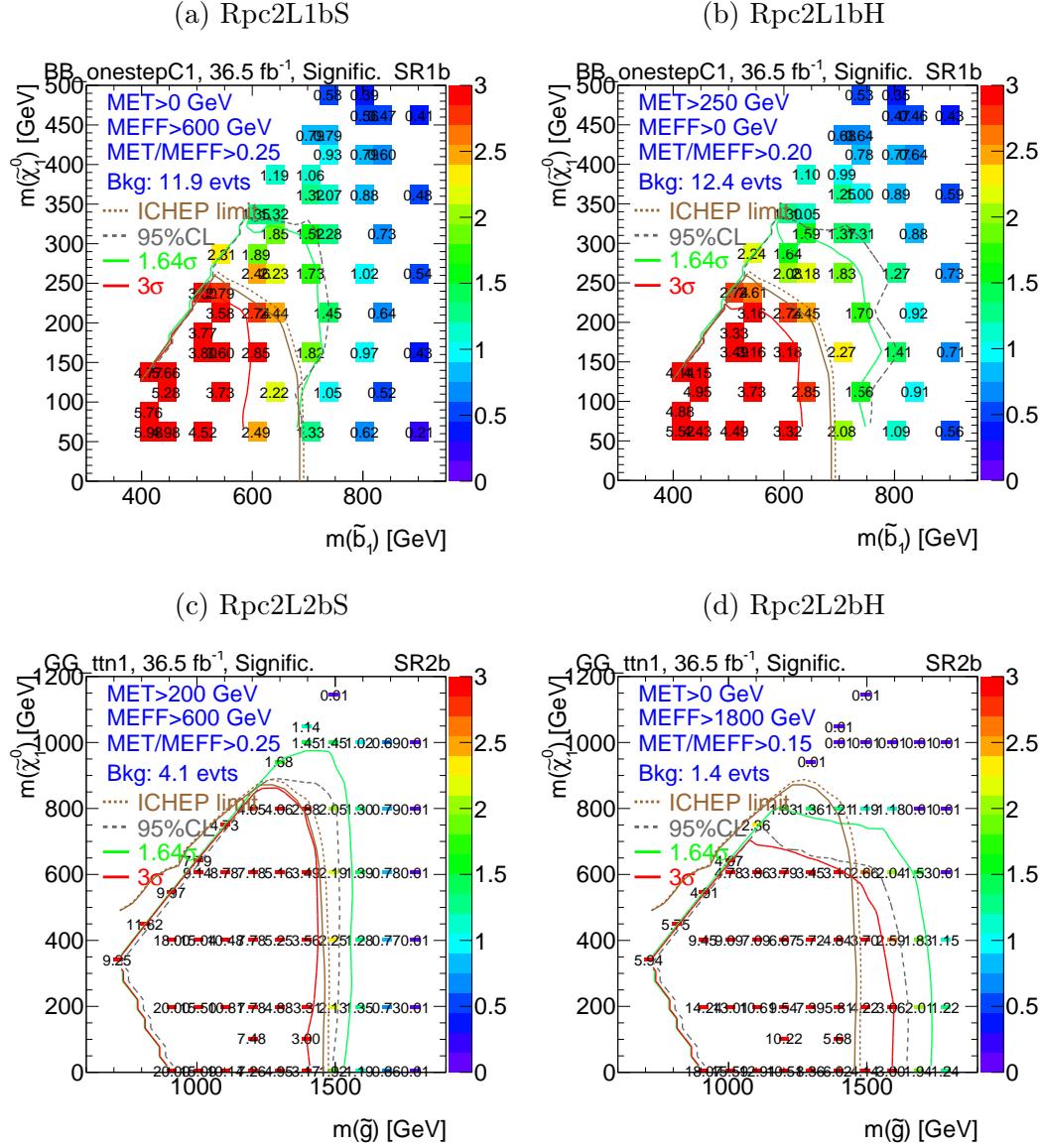
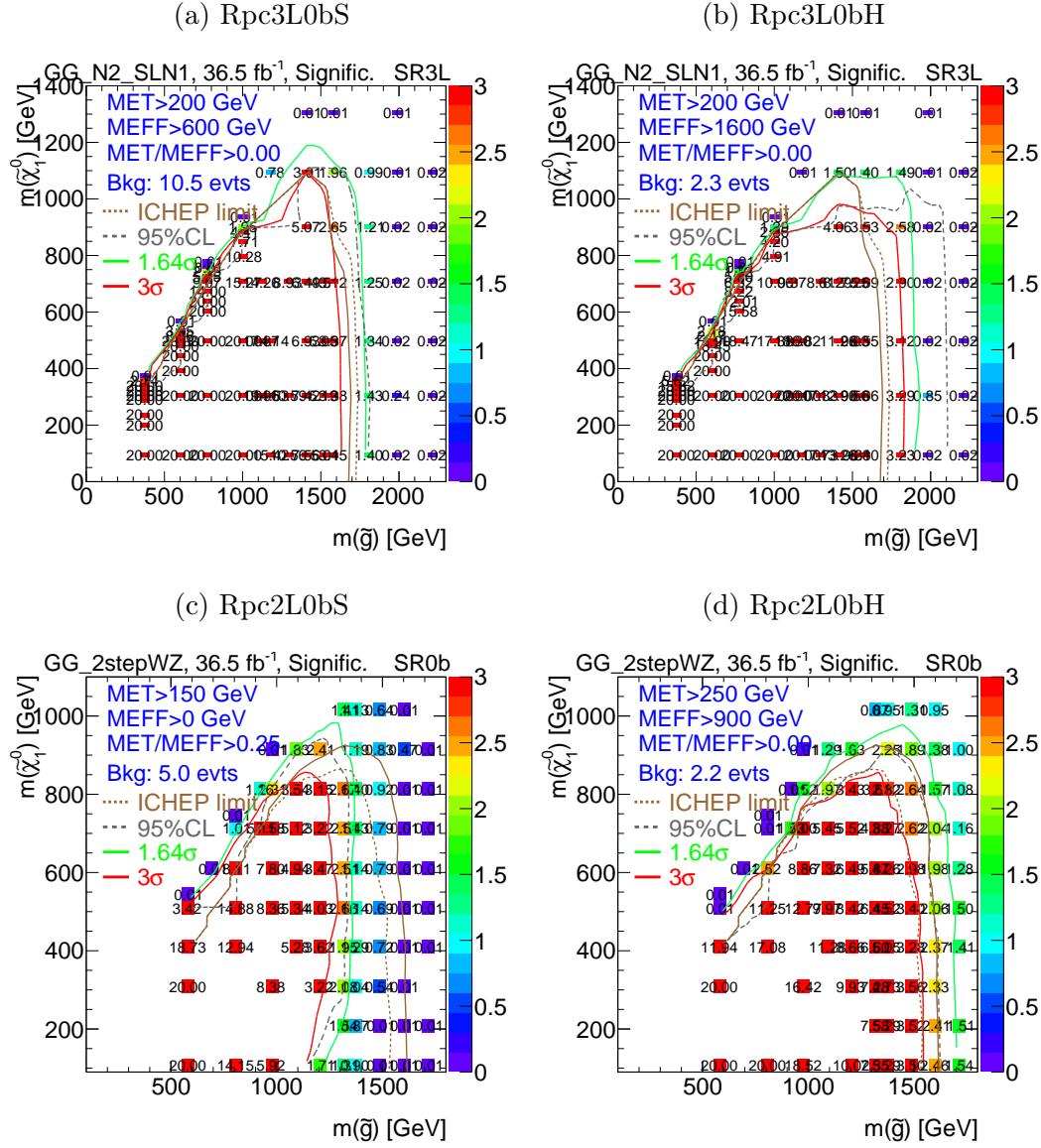


FIGURE 5.7: Discovery significance for the SRs with b -jets defined in Table 5.6 for 36.5 fb^{-1} : Rpc2L1bS and Rpc2L1bH in the $\tilde{b}_1\tilde{b}_1^* \rightarrow tt\tilde{\chi}_1^+\tilde{\chi}_1^-$ grid (top), Rpc2L2bS and Rpc2L2bH in the $\tilde{g}\tilde{g} \rightarrow tt\bar{t}\tilde{\chi}_1^0\tilde{\chi}_1^0$ grid (bottom). The 95% CL, 1.64σ , and 3σ discovery contours from the proposed signal regions are shown in grey, green, and red, respectively.

FIGURE 5.8: Discovery significance for the SRs without b -jets defined in Table 5.6

for 36.5 fb^{-1} : Rpc2L0bS and Rpc2L0bH in the $\tilde{g}\tilde{g}$ with $\tilde{g} \rightarrow q\bar{q}'WZ\tilde{\chi}_1^0$ grid (top)

and Rpc3L0bS and Rpc3L0bH in the $\tilde{g}\tilde{g}$ with $\tilde{g} \rightarrow q\bar{q}(\ell\ell/\ell\nu)\tilde{\chi}_1^0$ grid (bottom).

The 95% CL, 1.64σ , and 3σ discovery contours from the proposed signal regions are shown in grey, green, and red, respectively.

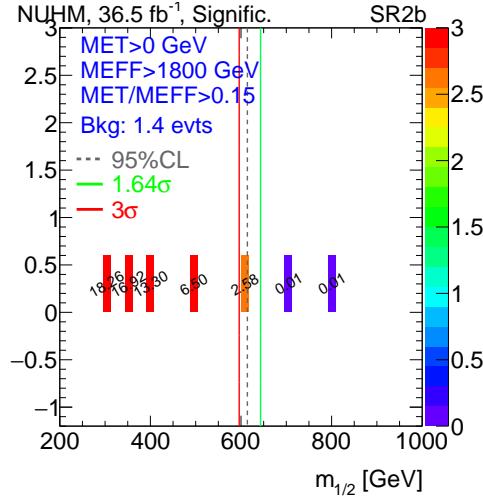


FIGURE 5.9: Discovery significance for Rpc2L2bH signal region for 36.5 fb^{-1} , NUHM2 model. The 95% CL, 1.64σ , and 3σ discovery contours from the proposed signal regions are shown in grey, green, and red, respectively.

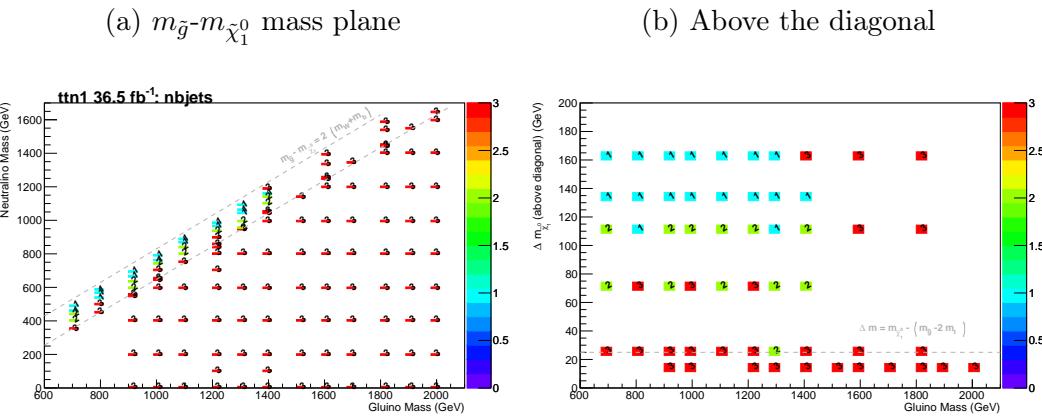


FIGURE 5.10: Optimal cut on the number of b -jets leading to the best discovery significance.

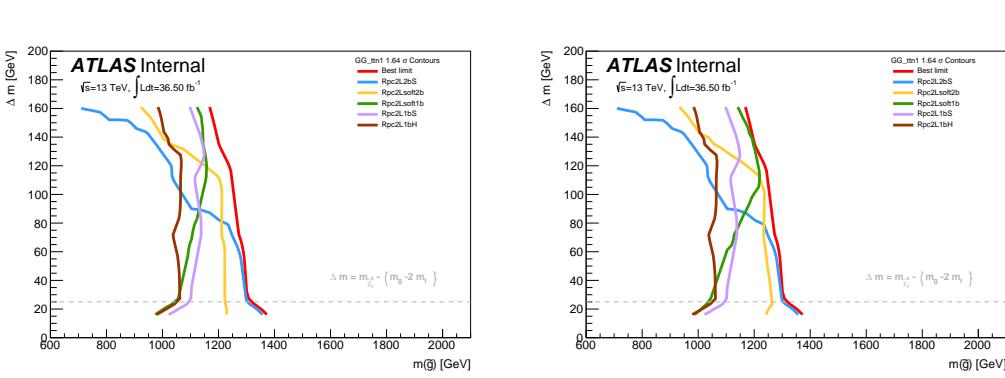


FIGURE 5.11: Comparison of significance contours at 1.64σ for 36.5 fb^{-1} between Rpc2Lsoft2b and Rpc2Lsoft1b and other signal regions in the off-diagonal region considering the option of an upper cut on the leading lepton p_T .

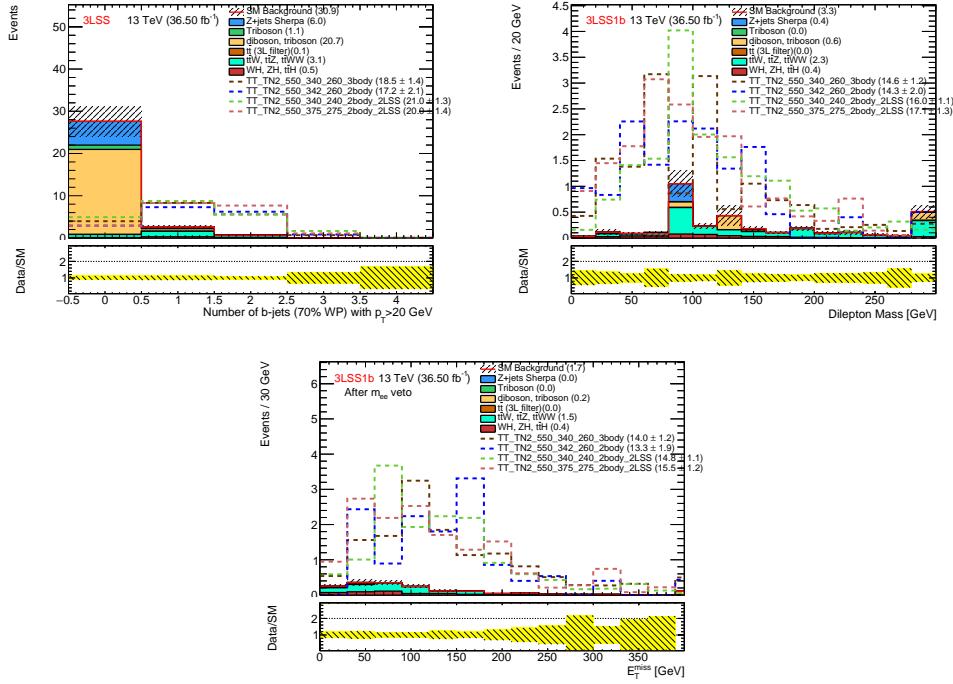


FIGURE 5.12: b -jet multiplicity after a 3LSS selection (top left), dilepton invariant mass distributions after a 3LSS plus ≥ 1 b -jet selection (top right), and E_T^{miss} distribution after a 3LSS, ≥ 1 b -jet and $81 < m_{e^\pm e^\pm} < 101$ GeV veto selection (bottom), all corresponding to 36.5 fb^{-1} . The background distributions are stacked, while the lines show the predictions for four signal points at \tilde{t} mass of 550 GeV.

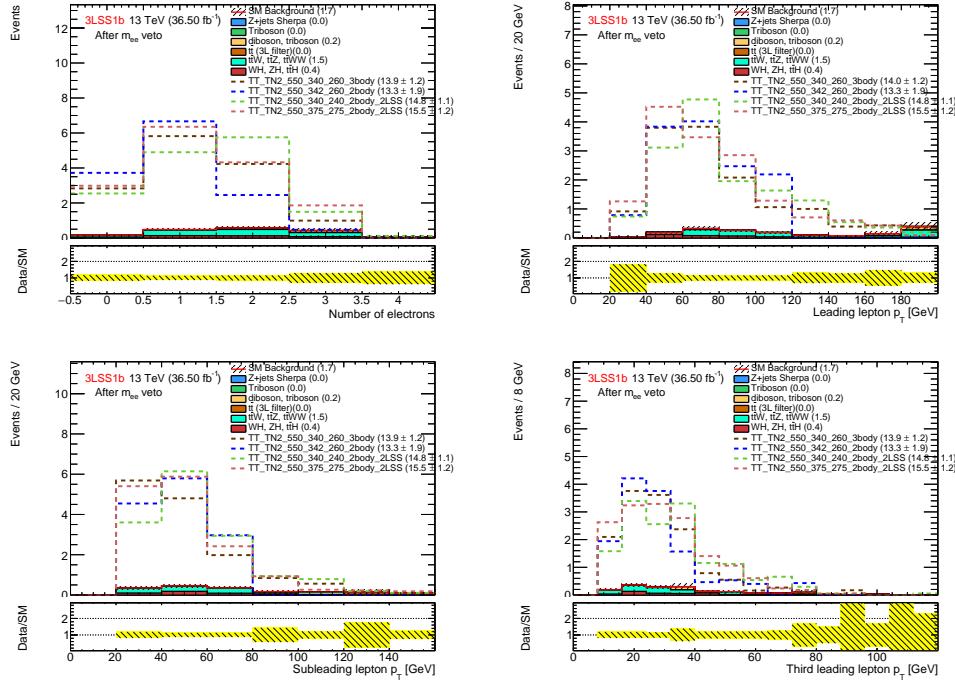


FIGURE 5.13: Number of electron (top left), and p_T of the leading (top right), subleading (bottom left) and third leading lepton (bottom right) after a 3LSS, ≥ 1 b -jet and $81 < m_{e^\pm e^\pm} < 101$ GeV veto selection (bottom), all corresponding to 36.5 fb^{-1} . The background distributions are stacked, while the lines show the predictions for four signal points at \tilde{t} mass of 550 GeV.

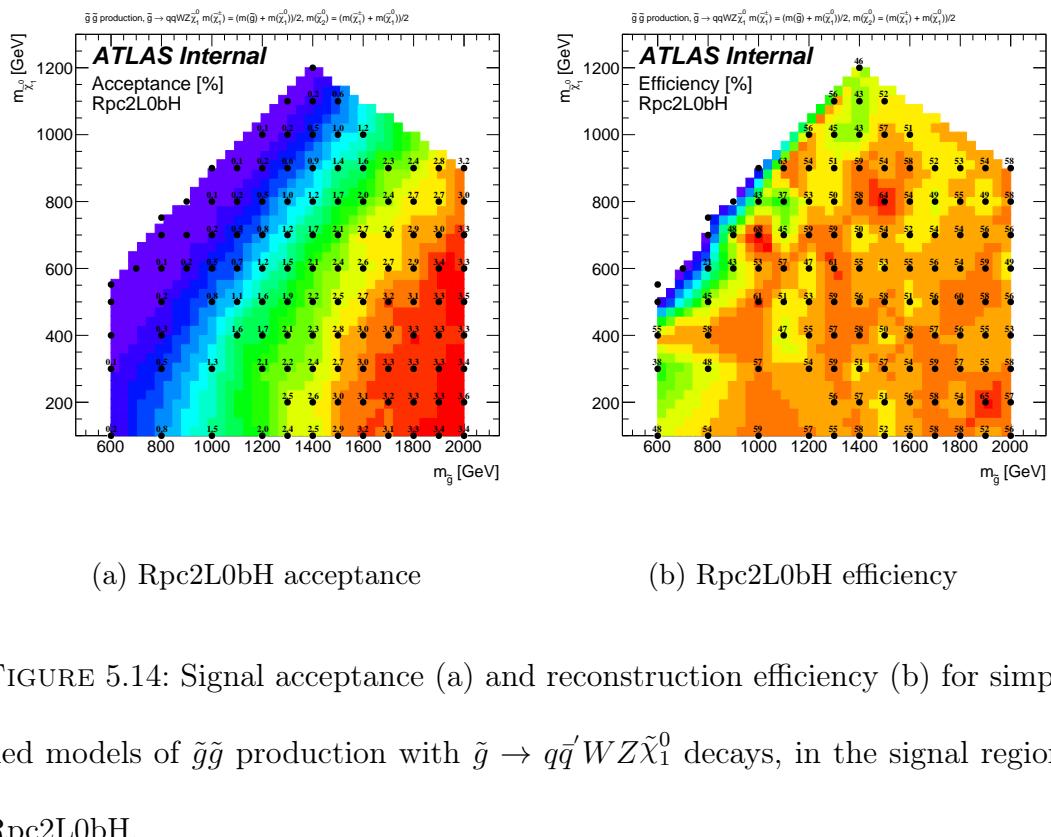


FIGURE 5.14: Signal acceptance (a) and reconstruction efficiency (b) for simplified models of $\tilde{g}\tilde{g}$ production with $\tilde{g} \rightarrow q\bar{q}'WZ\tilde{\chi}_1^0$ decays, in the signal regions Rpc2L0bH.

Chapitre 6

Data-driven Background Estimation Techniques

6.1 The problem of fakes

The reconstructed objects (leptons, photons, b -jets, etc.) in a collision event are used to perform a wide range of SM measurements or searches for evidence of BSM physics. The assumption is that these objects are ‘real’ representing the desired particles in the final state used in the analysis. In practice, the reconstructed objects might not be always ‘real’. In fact, they may be something completely different that was mistakenly reconstructed as the desired object, called ‘fake’. For the purpose of the analysis presented in this thesis, the focus is on ‘fake’ leptons. To illustrate the problem, a hadronic jet may deposit more energy in the electromagnetic calorimeter than the hadronic calorimeter, or that it leaves a narrow deposit of energy leading the reconstruction algorithms to mistake this jet for an electron. From the analysis point of view, the ‘fake’ electron will pass all the selection criteria and will be indistinguishable from a ‘real’ electron. It is important for the analysis that requires a reconstructed electron to model the fake electron background to get a sound result. This example was given with electrons, but can be generalized to muons as well. In short, any analysis that uses leptons in the final state must account for the ‘fake’ lepton background. This background can be more or less important depending on the detector, the analysis selection, and the number of leptons required. To estimate this background it is important

to first understand what type of processes lead to fake leptons.

6.2 Common processes for faking leptons

The reconstruction of ‘fake’ leptons can be an instrumental effect related to the inability to identify the object based on its measured properties by the detector. In this case, the reconstructed lepton is not a real lepton and the production process will be different for electrons and muons.

The reconstruction of electrons relies on the observation of well aligned particle hits in the layers of the ID that are consistent with an energy deposition in the EM calorimeter. Photons can mimic this signature since they deposit energy in the EM calorimeter that happens to be aligned with a charged track. A jet for example containing charged and neutral pions can lead to such scenarios. It is possible for the jet to have one charged pion leaving a track similar to that of an electron. The decay of π^0 mesons to photons in this jet can deposit energy in the EM calorimeter leading to the required signature. Another mechanism that can lead to fake electrons is the emission of photons via Bremsstrahlung from high energy muons. The muon track can be mistaken for that of an electron and the photons interact with the EM calorimeter leading to a signature similar to that of electrons. An additional process is that of photon conversions into a e^+e^- .

The reconstruction of muons relies on the observation of tracks from the ID matched to tracks from the muon spectrometer. It is possible for charged hadrons with long lifetime to traverse the calorimeter layers and leave hits in the muon

spectrometer. These hits may coincide with other hits from the ID due to the random activity in the event. As a result, a muon can get reconstructed. Another instance may occur when pions or kaons decay in-flight to muons in the muon spectrometer and happen to align with the primary vertex.

The leptons that are used in the physics analyses must be coming from the hard scatter, generally referred to as prompt leptons. There is another case where the reconstructed lepton is a real lepton but is not a lepton coming from the hard interaction, referred to as non-prompt leptons. Non-prompt leptons can be produced from heavy flavor meson decays with a low energy activity around the lepton which allows it to pass isolation requirements. A good example of this type of process is the semi-leptonic decay of top quark pair which contribute to final states with two leptons.

For the rest of the thesis, the fake leptons will be referred to as fake/non-prompt (FNP) leptons. There are several methods used to perform the estimation of FNP lepton backgrounds. A method that the author developed will be described next along with a standard method for estimating this type of backgrounds. The benefit of having two methods for estimating the FNP lepton background is to have enough confidence in the final estimate. The two methods use different assumptions which naturally leads to a more robust estimation of this difficult background. Moreover, the final estimate of the FNP lepton background is taken as a statistical combination of the estimates from the two methods leading to a reduction of the systematic uncertainties on the estimate.

6.3 The Monte Carlo Template Method

6.3.1 Motivation

The processes leading to FNP leptons depend on the selection applied in the analysis. For instance, a selection with same-sign leptons will have contributions from top quark pair production ($t\bar{t}$) or the associated production of a vector boson and jets ($W+jets$ or $Z+jets$). These processes cannot give two leptons of the same electric charge unless there is a charge mis-measurement (mainly affecting electrons) or that a FNP lepton was produced. It is possible to generate the processes that can contribute to a FNP lepton, such as $t\bar{t}$ or $V+jets$, with Monte Carlo event generators processed through Geant4 detector simulation of the ATLAS detector. This approach will yield an estimate however it might not be reliable. For instance, the detector simulation itself might not reproduce the true behavior of the interaction of the physics objects with the detector, particularly when looking at rare processes such as the production of FNP leptons. The second limitation is in the generation of enough MC events to probe the region of the phase space targeted by the analysis which affects the statistical uncertainties in the estimates. The latter concern is addressed by ensuring that the simulations for the major backgrounds ($t\bar{t}$ and $V+jets$) have much higher event count than the corresponding number of events observed in the data sample. In fact, these backgrounds have a large number of simulated events because they are important for many analyses (including SM measurements and BSM searches). The rest of

the section will concentrate on addressing the former limitation.

6.3.2 Description of the method

The MC template method relies on the correct modelling of FNP leptons kinematics in MC simulation to extrapolate background predictions from control regions to the signal regions. The method assumes that the kinematic shapes for each source of FNP lepton is correctly modeled in the simulations, and the normalization for each source is extracted in a combined fit to data control regions. The number of normalization factors depend on the number of identified origins of the FNP lepton in the signal regions and the control regions are designed to constrain these factors in regions enriched with FNP leptons from the same origin.

To illustrate the approach, we describe the application of the method in SS/3L analysis later described in this thesis. The processes of interest that may lead to a FNP lepton or a charge flip are $t\bar{t}$ and $V+\text{jets}$. FNP leptons are classified using an algorithm that navigates the generator particle record to determine where the FNP lepton is originating from. The lepton is classified as either an electron or a muon that is prompt from decays of on-shell W and Z bosons, non-prompt from a heavy flavor b decay (HF), or fake from mis-identification of a light flavor jet or a photon (LF). In the case of an electron, we further classify the prompt electrons to prompt electrons with the correct charge or with a charge mis-measurement, commonly named charge flip. In total, five categories referred to as MC templates are construed following the classification illustrated in figure 6.1.

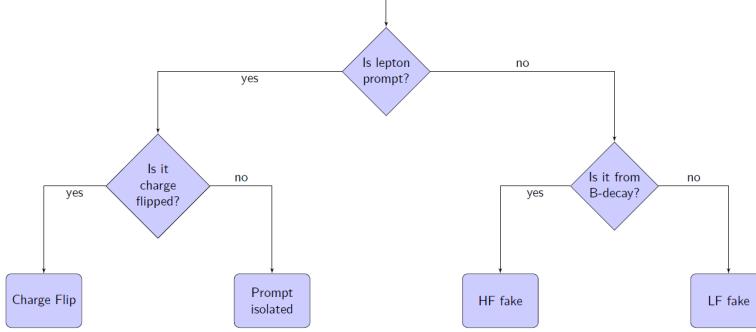


FIGURE 6.1: Lepton classification.

6.3.3 Correction factors

The FNP estimate relies on kinematic extrapolation using processes expected to contribute via FNP leptons from control regions with low jet multiplicity and E_T^{miss} , to the signal regions that require high jet multiplicity and E_T^{miss} . The control regions are chosen to separate FNP leptons from HF origins and FNP leptons from LF origins. For instance, a control sample characterized by the presence of a b -jet will be enriched in processes with one FNP lepton that is coming from a HF decay, while a sample characterized by the absence of a b -jet will have one FNP lepton from LF decay. The presence of one FNP lepton in the control sample allows the correction of the production rate of these FNP leptons by performing a fit to data.

For example, if a $Z \rightarrow \mu\mu + \text{LF jet}$ event is reconstructed as a $\mu^+\mu^-e^+$ event, then the electron is fake. Therefore, a correction of $\text{LF jet} \rightarrow e$ ($\text{Fr}(\text{LF} \rightarrow e)$) is applied to the rate of $\mu\mu e$ events. The correction $\text{Fr}(\text{LF} \rightarrow e)$ is constrained by a fit to data in control regions dominated by LF jet $\rightarrow e$ type fakes. Similarly, three other corrections are defined as LF jet $\rightarrow \mu$ ($\text{Fr}(\text{LF} \rightarrow \mu)$), HF jet $\rightarrow e$

(Fr(HF→e)), HF jet → μ (Fr(HF→ μ)). An additional correction is applied to correct the charge flip rate predicted by simulation. For example, a $Z \rightarrow e^+e^-$ event is reconstructed as e^+e^+ or e^-e^- . The simulation takes into account the charge flip rate but it might be off. The charge flip (Cf(e)) correction derived from a data fit is expected to recover this mis-modeling. The charge flip rate only concern electrons as the muon charge flip rate is negligible.

A likelihood fit is defined as the product of the Poisson probabilities describing the observed events in the binned distributions from the expected number of events rescaled by the five multipliers which are left free to float in the fit. These multipliers are applied to the MC predictions in the signal regions to obtain an estimation of the charge flip and FNP backgrounds.

6.3.4 Control regions

The corrections depend on the simulated sample, the reconstructed final state, and the flavor of the leptons. As a result, care must be taken when designing the control regions used to perform the fit of the FNP leptons and electron charge flip templates. For instance, each template needs to be constrained in a selection that is representative of the processes leading to FNP leptons and charge flip electrons present in the kinematic region targeted by the search for BSM physics.

In the SS/3L analysis discussed in this thesis the control regions are defined with at least two same-sign leptons, $E_T^{\text{miss}} > 40$ GeV, two or more jets. This pre-selection ensures that the FNP leptons are not from fakes originating from QCD like event topologies. They are further split in regions with or without b -jets

to constrain the HF and LF leptons respectively. In addition, they are also split with different flavours of the same-sign lepton pair ee, e μ , and $\mu\mu$, giving a total of six control regions. Any event entering the signal region is vetoed. The ee channel will constrain the charge flip correction factor, fake leptons from LF decays in the selection without b -jets, and non-prompt decay from HF in the selection with b -jets. The $\mu\mu$ channel will constrain the muon fake rates in the LF and HF decays for the selection without or with b -jets, respectively. The e μ channel will constrain both the electron and muon fakes for events containing both lepton flavors.

The six distributions are chosen for variables that provide the best separation between processes with prompt leptons and processes with FNP leptons and charge flip and are shown before and after the fit in Figures 6.2-6.4 and Figures 6.3-6.5, respectively.

The minimization of the negative log likelihood using the MINUIT package leads to the multipliers shown in Tables 6.1 and 6.2. The tables represent the multipliers obtained from the fit upon using two different parton showers, POWHEG-Box and SHERPA for the processes that lead to FNP leptons and charge flips. The systematic uncertainty is obtained by varying the generator from POWHEG-Box to SHERPA and evaluating the impact on the expected background from FNP and charge flip leptons. It is found to be the dominant contribution to the systematic uncertainty of the method (up to 80%). The uncertainties in the multipliers themselves correspond to how much the parameter needs to be varied for a one standard deviation change in the likelihood function. This uncertainty takes into account the limited number of simulated events and is included as a systematic

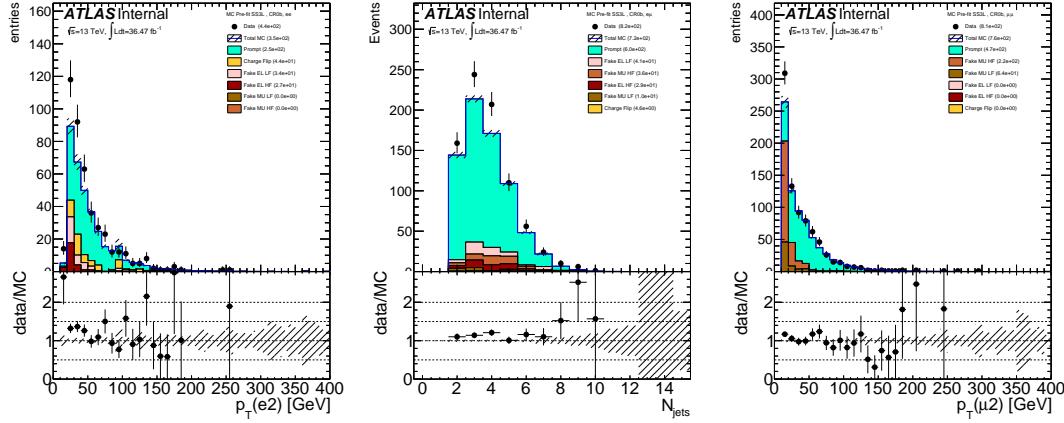


FIGURE 6.2: Pre-fit distributions for ee channel (left), for $e\mu$ channel (middle), and for $\mu\mu$ channel (right) from CR0b that were used in the fit to extract the FNP lepton and charge flip multipliers. The generator used in these plots is Powheg. The hashed band represents the sum of systematic uncertainties on the predictions.

uncertainty on the expected number of background events.

TABLE 6.1: The FNP and charge flip multipliers obtained after minimizing the likelihood function using Pythia. The uncertainty in the multipliers takes into account the limited statistics of simulated events.

Category	Multiplier	Uncertainty
chFlip	1.49	0.58
HF EL	2.80	0.98
LF EL	2.89	0.88
HF MU	1.59	0.31
LF MU	1.00	1.34

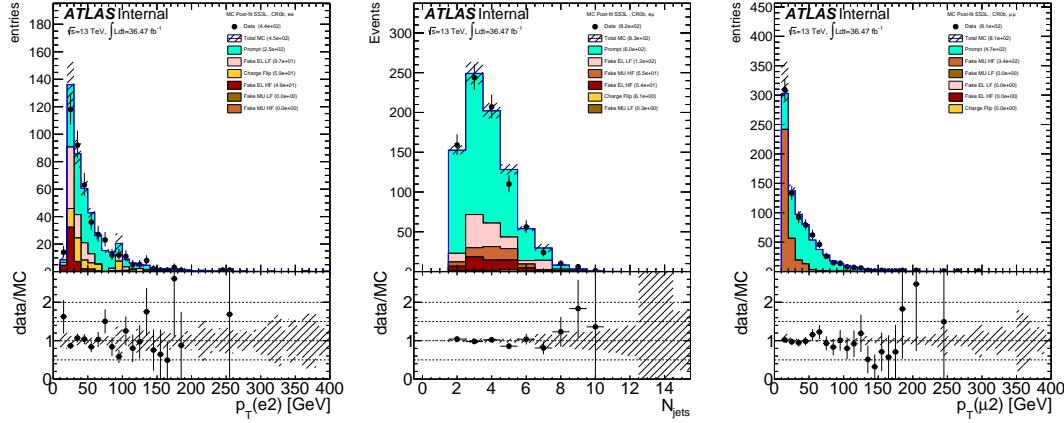


FIGURE 6.3: Post-fit distributions for ee channel (left), for $e\mu$ channel (middle), and for $\mu\mu$ channel (right) from CR0b that were used in the fit to extract the FNP lepton and charge flip multipliers. The generator used in these plots is Powheg. The hashed band represents the sum of systematic uncertainties on the predictions.

TABLE 6.2: The FNP and charge flip multipliers obtained after minimizing the likelihood function using Sherpa. The uncertainty in the multipliers takes into account the limited statistics of simulated events.

Category	Multiplier	Uncertainty
chFlip	1.34	0.58
HF EL	2.40	0.85
LF EL	1.83	1.04
HF MU	1.17	0.16
LF MU	2.40	0.81

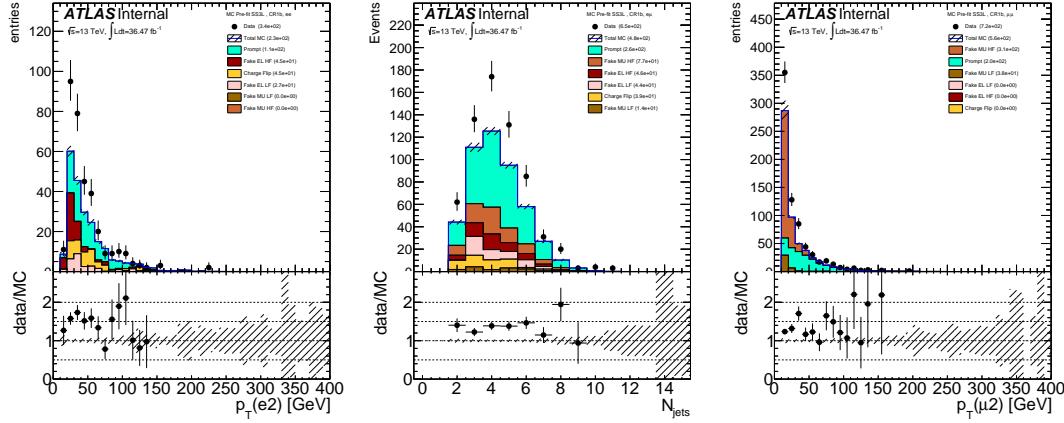


FIGURE 6.4: Pre-fit distributions for ee channel (left), for $e\mu$ channel (middle), and for $\mu\mu$ channel (right) from CR1b that were used in the fit to extract the FNP lepton and charge flip multipliers. The generator used in these plots is Powheg. The hashed band represents the sum of systematic uncertainties on the predictions.

6.4 Matrix Method

The FNP leptons do not often pass one of the lepton selection criteria but have non-zero impact parameter, and are often not well-isolated. These selection requirements are key ingredients to control the FNP leptons. The number of events with at least one FNP lepton is estimated using two classes of leptons : a real-enriched class of “tight” leptons corresponding to signal leptons and a fake-enriched class of “loose” leptons corresponding to candidate leptons with

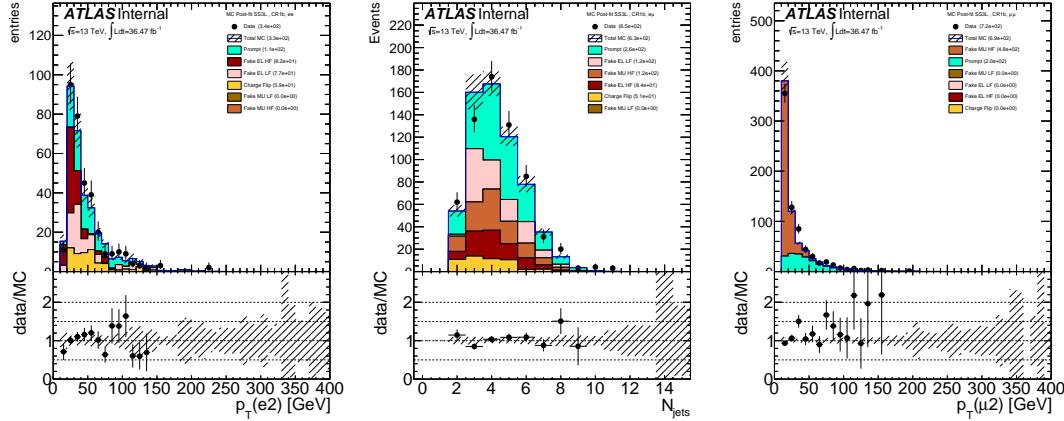


FIGURE 6.5: Post-fit distributions for ee channel (left), for $e\mu$ channel (middle), and for $\mu\mu$ channel (right) from CR1b that were used in the fit to extract the FNP lepton and charge flip multipliers. The generator used in these plots is Powheg. The hashed band represents the sum of systematic uncertainties on the predictions.

relaxed identification criteria¹. In the next sections, a description of the simplest form of the matrix method will be given with events containing one object. Then a generalized treatment that can handle events with an arbitrary number of leptons in the final states will be discussed.

6.4.1 Events with one object

Given the probabilities ε/ζ for a real/FNP candidate lepton to satisfy the signal lepton criteria, one can relate the number of events with one candidate lepton passing/failing signal requirements ($n_{\text{pass}}/n_{\text{fail}}$) to the number of events

1. Signal leptons are leptons satisfying the signal lepton definition, while the candidate leptons are leptons satisfying some pre-selection cuts and usually passing the overlap removal requirements as discussed in the analysis Section 5.4.3.

with one real/FNP signal leptons ($n_{\text{real}}/n_{\text{FNP}}$) :

$$\begin{pmatrix} n_{\text{pass}} \\ n_{\text{fail}} \end{pmatrix} = \begin{pmatrix} \varepsilon & \zeta \\ 1 - \varepsilon & 1 - \zeta \end{pmatrix} \begin{pmatrix} n_{\text{real}} \\ n_{\text{FNP}} \end{pmatrix}; \quad (6.1)$$

allowing to determine the unknown number of events n_{FNP} from the observed n_{pass} and n_{fail} given measurements of the probabilities ε/ζ .

The predictive power of the matrix method comes from the fact that the real and FNP leptons have different composition in the two collections of tight and loose objects leading to $\varepsilon \neq \zeta$. In fact, the tight lepton collection will be dominated by real objects while the loose region will be dominated by fake objects. As a result, the inequality $\varepsilon \gg \zeta$ will always hold true which guarantees that the matrix in Eq. 6.1 is invertible and gives positive estimates.

The next step is to invert the relation in Eq. 6.1 to obtain

$$\begin{pmatrix} n_{\text{real}} \\ n_{\text{FNP}} \end{pmatrix} = \frac{1}{\varepsilon - \zeta} \begin{pmatrix} \bar{\zeta} & -\zeta \\ -\bar{\varepsilon} & \varepsilon \end{pmatrix} \begin{pmatrix} n_{\text{pass}} \\ n_{\text{fail}} \end{pmatrix}; \quad (6.2)$$

where $\bar{\varepsilon} = 1 - \varepsilon$ and $\bar{\zeta} = 1 - \zeta$. The FNP lepton component is :

$$n_{\text{FNP}} = \frac{1}{\varepsilon - \zeta} ((\varepsilon - 1) n_{\text{pass}} + n_{\text{fail}}). \quad (6.3)$$

However, the quantity of interest is the expected FNP lepton background that passes the tight selection criteria : $n_{\text{pass}} \cap \text{FNP} = \zeta n_{\text{FNP}}$. To obtain this quantity, the identity from Eq. 6.1 is used to get :

$$n_{\text{FNP}} = \frac{\zeta}{\varepsilon - \zeta} ((\varepsilon - 1) n_{\text{pass}} + n_{\text{fail}}). \quad (6.4)$$

The linearity of Eq. 6.4 with respect to n_{pass} and n_{fail} allows the method to be applied on an event-by-event, effectively resulting into a weight being assigned to each event. By defining

$$n_{\text{pass}} = \sum_{\text{all events}} \mathbb{1}_{\text{pass}}, \quad n_{\text{fail}} = \sum_{\text{all events}} \mathbb{1}_{\text{fail}}, \quad \mathbb{1}_{\text{fail}} = 1 - \mathbb{1}_{\text{pass}},$$

where $\mathbb{1}_{\text{pass(fail)}} = 1$ if the object pass (fail) the tight selection requirement and $\mathbb{1}_{\text{pass(fail)}} = 0$ otherwise. Eq. 6.4 can be written as

$$n_{\text{FNP}} = \sum_{\text{all events}} \left\{ \frac{\zeta}{\varepsilon - \zeta} (\varepsilon - \mathbb{1}_{\text{pass}}) \right\} = \sum_{\text{all events}} \omega$$

where

$$\omega = \frac{\zeta}{\varepsilon - \zeta} (\varepsilon - \mathbb{1}_{\text{pass}}) \quad (6.5)$$

is the weight to be assigned to each event in the case of one FNP lepton in the event. The generalization of this formalism to higher dimensions with multiple objects will be covered next.

6.4.2 Dynamic matrix method

The one lepton case readily generalizes to events with more than one lepton in a formalism that can handle an arbitrary number of leptons in the event. The method should be applied event-by-event, effectively resulting into a weight being

assigned to each event. The predicted yield of events with FNP leptons is simply the sum of weights. A general formula will be derived starting from the two objects case then specific examples will be given to illustrate the application of the method.

If two objects are present in the event, the probabilities ε/ζ will depend on the kinematic properties of these objects. Typically the probables will vary as a function of p_T and $|\eta|$. For this reason, the probabilities will be different and will have an index to identify the object under study : ε_i/ζ_i where $i = 1, 2, \dots$. An identity similar to Eq. 6.1 can be formed for two objects with a change in notation for simplicity :

$$\begin{pmatrix} N_{TT} \\ N_{TL} \\ N_{LT} \\ N_{LL} \end{pmatrix} = \Lambda \times \begin{pmatrix} N_{RR} \\ N_{RF} \\ N_{FR} \\ N_{FF} \end{pmatrix}, \quad (6.6)$$

where $(N_{RR}, N_{RF}, N_{FR}, N_{FF})$ are the number of events with respectively two real, one real plus one FNP (two terms), and two FNP leptons before applying tight cuts, respectively, and $(N_{TT}, N_{TL}, N_{LT}, N_{LL})$ are the observed number of events for which respectively both lepton pass the tight cut, only one of them (two terms), or both fail the tight cut, respectively.

Λ is given by :

$$\Lambda = \begin{pmatrix} \varepsilon_1 \varepsilon_2 & \varepsilon_1 \zeta_2 & \zeta_1 \varepsilon_2 & \zeta_1 \zeta_2 \\ \varepsilon_1 (1 - \varepsilon_2) & \varepsilon_1 (1 - \zeta_2) & \zeta_1 (1 - \varepsilon_2) & \zeta_1 (1 - \zeta_2) \\ (1 - \varepsilon_1) \varepsilon_2 & (1 - \varepsilon_1) \zeta_2 & (1 - \zeta_1) \varepsilon_2 & (1 - \zeta_1) \zeta_2 \\ (1 - \varepsilon_1) (1 - \varepsilon_2) & (1 - \varepsilon_1) (1 - \zeta_2) & (1 - \zeta_1) (1 - \varepsilon_2) & (1 - \zeta_1) (1 - \zeta_2) \end{pmatrix}$$

which can also be written in terms of a Kronecker product in Eq. 6.6 to obtain :

$$\begin{pmatrix} N_{TT} \\ N_{TL} \\ N_{LT} \\ N_{LL} \end{pmatrix} = \begin{pmatrix} \varepsilon_1 & \zeta_1 \\ \bar{\varepsilon}_1 & \bar{\zeta}_1 \end{pmatrix} \otimes \begin{pmatrix} \varepsilon_2 & \zeta_2 \\ \bar{\varepsilon}_2 & \bar{\zeta}_2 \end{pmatrix} \begin{pmatrix} N_{RR} \\ N_{RF} \\ N_{FR} \\ N_{FF} \end{pmatrix} \quad (6.7)$$

To make the notation more compact, the set of 4 numbers $(N_{TT}, N_{TL}, N_{LT}, N_{LL})$ can be represented by a rank 2 tensor $\mathcal{T}_{\alpha_1 \alpha_2}$ where α_i corresponds to one object that is either tight (T) or loose (L). Similarly the numbers $(N_{RR}, N_{RF}, N_{FR}, N_{FF})$ can be represented by $\mathcal{R}_{\alpha_1 \alpha_2}$ where α_i corresponds to one object that is either real (R) or FNP (F). With this convention, the Kronecker product of Eq. 6.7 can be obtained by contracting each index α_i of the tensors \mathcal{T} or \mathcal{R} by the 2×2 matrix

$$\phi_{i \beta_i}^{\alpha_i} :$$

$$\mathcal{T}_{\beta_1 \beta_2} = \phi_{1 \beta_1}^{\alpha_1} \phi_{2 \beta_2}^{\alpha_2} \mathcal{R}_{\alpha_1 \alpha_2}, \quad \phi_i = \begin{pmatrix} \varepsilon_i & \zeta_i \\ \bar{\varepsilon}_i & \bar{\zeta}_i \end{pmatrix} \quad (6.8)$$

Following the same procedure as in the one object case, the matrix inversion of the 4×4 λ matrix is simplified to a matrix inversion of the 2×2 ϕ matrices. The

quantity of interest is the FNP lepton background that passes the tight selection criteria as in Eq. 6.4 which can be compactly written in the two objects case as :

$$\mathcal{T}_{\nu_1 \nu_2}^{\text{FNP}} = \phi_{\nu_1}{}^{\mu_1} \phi_{\nu_2}{}^{\mu_2} \xi^{\beta_1 \beta_2}_{\mu_1 \mu_2} \phi^{-1}{}^{\alpha_1}_{\beta_1} \phi^{-1}{}^{\alpha_2}_{\beta_2} \mathcal{T}_{\alpha_1 \alpha_2}. \quad (6.9)$$

The tensor ξ encodes the component of tight and FNP lepton background. In the two objects case, ξ needs to select the total background with at least one fake lepton $N_F = N_{RF} + N_{FR} + N_{FF}$ that are also passing the tight selection criteria corresponding to the region with signal leptons. As a result, ξ takes the form :

$$\xi = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

To further illustrate, Eq. 6.9 can be written explicitly in the notation of Eq. 6.6 as :

$$N_{\text{FNP}}^{\text{signal}} = \begin{pmatrix} 0 & \varepsilon_1 \zeta_2 & \zeta_1 \varepsilon_2 & \zeta_1 \zeta_2 \end{pmatrix} \Lambda^{-1} \begin{pmatrix} N_{TT} \\ N_{TL} \\ N_{LT} \\ N_{LL} \end{pmatrix}$$

The generalization of Eq. 6.9 from the two objects case to m number of objects in the final state is straightforward :

$$\mathcal{T}_{\nu_1 \dots \nu_m}^{\text{FNP}} = \phi_{\nu_1}{}^{\mu_1} \cdots \phi_{\nu_m}{}^{\mu_m} \xi_{\mu_1 \dots \mu_m}^{\beta_1 \dots \beta_m} \phi^{-1}{}_{\beta_1}{}^{\alpha_1} \cdots \phi^{-1}{}_{\beta_m}{}^{\alpha_m} \mathcal{T}_{\alpha_1 \dots \alpha_m}. \quad (6.10)$$

The tensor ξ is of the general form

$$\xi_{\mu_1 \dots \mu_m}^{\beta_1 \dots \beta_m} = \delta_{\mu_1}^{\beta_1} \cdots \delta_{\mu_m}^{\beta_m} h(\beta_1, \dots, \beta_m, \nu_1, \dots, \nu_m)$$

where the function h can take values 0 or 1 based on the tight or loose configuration being computed which is encoded in the dependence on the indices ν_i .

The application of the matrix method to multilepton final states comes with two important remarks. Firstly, contributions of events with charge-flip electrons would bias a straightforward matrix method estimate (in particular for a final state formed by two leptons with same electric charge). This happens because the candidate-to-signal efficiency for such electrons is typically lower than for real electrons having a correctly-assigned charge. One therefore needs to subtract from n_{pass} and n_{fail} the estimated contributions from charge-flip. This can be performed by including as well events with pairs of opposite-sign candidate leptons in the matrix method estimate, but assigning them an extra weight corresponding to the charge-flip weight. Thanks (again) to the linearity of the matrix method with respect to n_{pass} and n_{fail} , this weight-based procedure is completely equivalent (but more practical) to the aforementioned subtraction.

Secondly, the analytic expression of the matrix method event weight depends on the lepton multiplicity of the final state. This concerns events with three or more candidate leptons : one such event takes part both in the evaluation of the FNP lepton background for a selection with two signal leptons or a selection with

three signal leptons, but with different weights². Therefore, for a given event used as input to the matrix method, one should consider all possible leptons combinations, each with its own weight and its own set of kinematic variables. For example, a $e^+e^-\mu^+$ event is used in the background estimate both as an $e^+\mu^+$ event (with a weight w_1) and as an $e^+e^-\mu^+$ event (with a weight $w_2 \neq w_1$).

6.4.3 Propagation of uncertainties

The two parameters (ε and ζ respectively) can be measured in data, and depend on the flavor and kinematics of the involved leptons. Systematic uncertainties resulting from the measurement of these two parameters, and their extrapolation to the signal regions, can be propagated to uncertainties on the event weight through standard first-order approximations. The different sources of uncertainties should be tracked separately so that correlations of uncertainties across different events can be accounted for correctly. The resulting set of uncertainties on the cumulated event weights can be then added in quadrature to form the systematic uncertainty on the predicted FNP lepton background yield. The corresponding statistical uncertainty can be taken as the RMS of the event weights.

2. This can appear for inclusive selections : for example an event with two signal leptons may or not contain additional candidate leptons, in a transparent way

Chapitre 7

Background Estimation

7.1 Overview

In this analysis, two types of backgrounds can be distinguished. The first category is the irreducible background from events with two same-sign prompt leptons or at least three prompt leptons and is estimated using the MC simulation samples (Section 7.2). Since diboson and $t\bar{t}V$ events are the main backgrounds in the signal regions, dedicated validation regions with an enhanced contribution from these processes, and small signal contamination, are defined to verify the background predictions from the simulation (Section 7.3). The second category is the reducible background, which includes events containing electrons with mis-measured charge, mainly from the production of top quark pairs, and events containing at least one fake or non-prompt (FNP) lepton. The application of the data-driven methods of Chapter 6 is presented in Section 7.2).

7.2 Irreducible Backgrounds

7.2.1 Expected yields in the signal regions

The predicted event yields in the signal regions are presented in Table 7.1, while the contributions of particular rare processes to the signal regions, relative to the summed contributions of all these processes, are shown in Table 7.2.

TABLE 7.1: Expected yields for background processes with prompt leptons, in the SRs proposed in Section 5.6, for 36.1 fb^{-1} . Quoted uncertainties include statistical sources only. Rare category includes $t\bar{t}WW$, $t\bar{t}WZ$, $3t$, tZ , tWZ , WH , ZH and VVV , and detailed contributions of these processes can be found in Table 7.2.

	$t\bar{t}V$	VV	$t\bar{t}H$	$t\bar{t}t\bar{t}$	rare
Rpc2L0bH	0.20 ± 0.05	1.14 ± 0.23	0.08 ± 0.04	0.02 ± 0.01	0.17 ± 0.04
Rpc2L0bS	0.82 ± 0.10	3.13 ± 0.21	0.26 ± 0.05	0.01 ± 0.00	0.20 ± 0.04
Rpc2L1bH	3.86 ± 0.20	0.61 ± 0.06	1.01 ± 0.10	0.53 ± 0.03	0.97 ± 0.12
Rpc2L1bS	3.94 ± 0.20	0.48 ± 0.05	1.28 ± 0.10	0.33 ± 0.03	0.87 ± 0.12
Rpc2L2bH	0.41 ± 0.05	0.04 ± 0.01	0.10 ± 0.03	0.17 ± 0.02	0.14 ± 0.04
Rpc2L2bS	1.57 ± 0.12	0.10 ± 0.03	0.44 ± 0.06	0.25 ± 0.02	0.32 ± 0.05
Rpc2Lsoft1b	1.24 ± 0.11	0.14 ± 0.02	0.44 ± 0.06	0.09 ± 0.01	0.18 ± 0.04
Rpc2Lsoft2b	1.15 ± 0.10	0.05 ± 0.02	0.37 ± 0.06	0.20 ± 0.02	0.17 ± 0.03
Rpc3L0bH	0.18 ± 0.04	2.64 ± 0.12	0.03 ± 0.02	0.01 ± 0.00	0.29 ± 0.04
Rpc3L0bS	0.99 ± 0.09	8.95 ± 0.21	0.12 ± 0.04	0.02 ± 0.01	0.75 ± 0.07
Rpc3L1bH	1.52 ± 0.11	0.48 ± 0.05	0.25 ± 0.06	0.28 ± 0.03	0.87 ± 0.12
Rpc3L1bS	7.02 ± 0.23	1.44 ± 0.10	1.36 ± 0.10	0.69 ± 0.04	2.51 ± 0.22
Rpc3LSS1b	0.00 ± 0.00	0.00 ± 0.00	0.21 ± 0.04	0.00 ± 0.00	0.09 ± 0.01

7.2.2 Validation regions

Dedicated validation regions are defined to verify the estimate of the $W^\pm W^\pm jj$, the $WZjjjj(j)$, the $t\bar{t}W$ and $t\bar{t}Z$ processes in the signal regions. For a better validation of WZ processes in association with a large jet multiplicity, two VRs are proposed : $WZ4j$ and $WZ5j$, with 4 and respectively 5 reconstructed jets in

TABLE 7.2: Contributions of particular rare processes to the signal regions, relative to the summed contributions of all these processes.

	VVV	VH	$3t$	tZ	$t\bar{t}WW$	tWZ	$t\bar{t}WZ$
Rpc2L0bH	23%	0%	2%	3%	25%	43%	1%
Rpc2L0bS	50%	0%	3%	15%	14%	16%	0%
Rpc2L1bH	2%	0%	7%	4%	41%	41%	2%
Rpc2L1bS	2%	0%	6%	3%	34%	50%	2%
Rpc2L2bH	3%	0%	15%	4%	47%	27%	1%
Rpc2L2bS	2%	0%	13%	2%	42%	36%	2%
Rpc2Lsoft1b	3%	0%	9%	0%	76%	7%	2%
Rpc2Lsoft2b	2%	0%	17%	4%	54%	19%	2%
Rpc3L0bH	52%	0%	0%	3%	1%	40%	1%
Rpc3L0bS	50%	0%	0%	4%	2%	39%	1%
Rpc3L1bH	3%	0%	3%	3%	17%	70%	1%
Rpc3L1bS	2%	0%	3%	7%	18%	64%	2%
Rpc3LSS1b	25%	0%	0%	0%	0%	0%	74%

the event. The corresponding selections are summarized in Table 7.3.

$W^\pm W^\pm + \text{jets validation region}$

The $W^\pm W^\pm + \text{jets}$ processes contribute mainly in the signal regions with no b -tagged jet requirement and two same-sign leptons. This validation region, $W^\pm W^\pm$ -VR, has exactly one SS pair (and no additional baseline leptons), zero b -jets and at least two jets with p_T above 50 GeV. Additional requirements on E_T^{miss} and m_{eff} help to decrease the amount of detector background as shown in Figure 7.2 (left

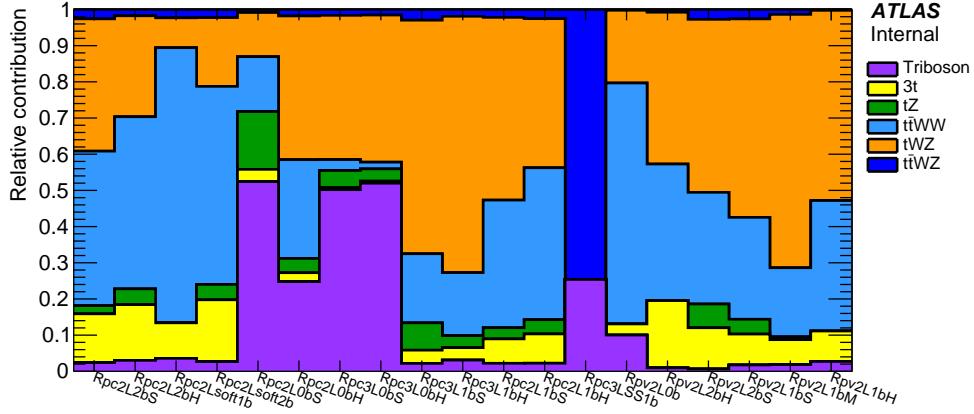


FIGURE 7.1: Relative contribution in each signal region from the processes in the category labelled as rare in the paper ($t\bar{t}WW$, $t\bar{t}WZ$, tZ , tWZ , $tt\bar{t}$, WH , ZH and triboson production).

and middle). To further improve the purity, the sub-leading lepton p_T is increased to 30 GeV, and cuts on minimum angular separation between the leptons and jets, and between the two leptons (Figure 7.2, right) are placed as detailed in Table 7.3. The purity is around 34% with this definition of the validation region. Signal contamination (highly reduced by applying a veto of all SRs) it is found to be at most 5% when looking at $\tilde{g} \rightarrow q\bar{q}(\tilde{\ell}\ell/\tilde{\nu}\nu)$ scenarios.

$WZ + \text{jets validation region}$

Contribution from $WZ+\text{jets}$ processes can be significant in regions vetoing the presence of b -jets and requiring three leptons. Given the large data sample collected, it is possible to design validation regions that require at least four and even at least five jets with $p_T > 25$ GeV in the event. Thus, two validation regions, $WZ4j\text{-VR}$ and $WZ5j\text{-VR}$, are proposed to better probe the modelling of the jet multiplicity in WZ processes. Both regions are defined with exactly three signal leptons and

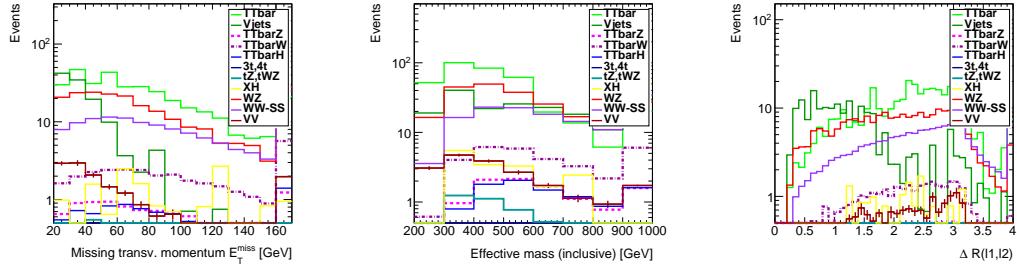


FIGURE 7.2: E_T^{miss} (left), m_{eff} (middle) and $\min \Delta R(\ell_1, \ell_2)$ (right) after lepton and jet selection of the $W^\pm W^\pm$ -VR (and no additional requirements). Signal regions are vetoed. All MC samples are normalized to a luminosity of 36.1 fb^{-1} . Last bin includes overflow.

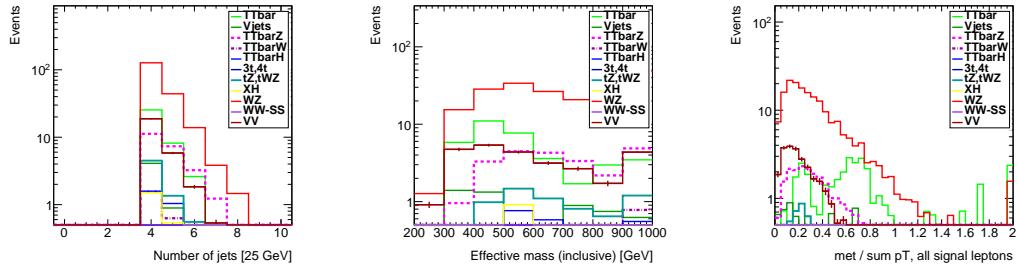


FIGURE 7.3: Number of jets with $p_T > 25 \text{ GeV}$ (left), m_{eff} (middle) and ratio between the E_T^{miss} in the event and the sum of all lepton p_T (right) after lepton and four jet selection of the WZ -VR (and no additional requirements). Signal regions are vetoed as detailed in Table 7.3. All MC samples are normalized to a luminosity of 36.1 fb^{-1} . Last bin includes overflow.

no fourth baseline lepton, to reduce the ZZ background contamination. The m_{eff} , and the upper cut on the ratio between the E_T^{miss} in the event and the sum of all lepton p_T are great discriminants against the reducible backgrounds. Some kinematic distributions after lepton and four jet selection are shown in Figure 7.3.

Purity in $WZ4j$ -VR ($WZ5j$ -VR) is around 67% (64%). When looking at

$\tilde{g} \rightarrow q\bar{q}(\tilde{\ell}\ell/\tilde{\nu}\nu)$ scenarios, the signal contamination is below 5% in most of the non-excluded phase space, except for small $\Delta M(\tilde{g}, \tilde{\chi}_1^0)$ where it can go up to 30% (15%). Signal contamination is found to be much lower for $\tilde{g} \rightarrow q\bar{q}WZ\tilde{\chi}_1^0$ scenarios.

$t\bar{t} + W$ background validation

A $t\bar{t} + W$ validation region ($t\bar{t}W$ -VR) is defined with exactly one SS lepton pair and at least one b -jet. At least four jets are required in the ee and $e\mu$ channels, while in the $\mu\mu$ channel the selection is relaxed to at least three jets (less reducible background); also the jet p_T thresholds are different between these two cases (same motivation). As shown in Figure 7.4, the amount of $t\bar{t}$ background after this pre-selection is still very large and additional requirements on E_T^{miss} , m_{eff} and on the ratio between the sum of p_T of all b -jets and the sum of p_T of all jets are placed as mentioned in Table 7.3. With this definition the achieved purity in $t\bar{t}W$ -VR is 33%. Signal contamination is around 20% when looking at $\tilde{b}\tilde{b}$ SUSY models.

$t\bar{t} + Z$ background validation

A $t\bar{t} + Z$ enriched validation region (ttZ -VR) is defined with at least one SFOS lepton pair and at least one b -jet. At least three jets are required in the event regardless of the lepton channel. Some kinematic distributions after this pre-selection are shown in Figure 7.5 (left and middle). To increase the purity, the invariant mass of the SFOS lepton pair is selected to be inside the [81, 101] GeV interval and the m_{eff} in the event should be greater than 450 GeV. With this selection, the purity is 58%. One can increase it even further (by $\sim 10\%$) if at least two

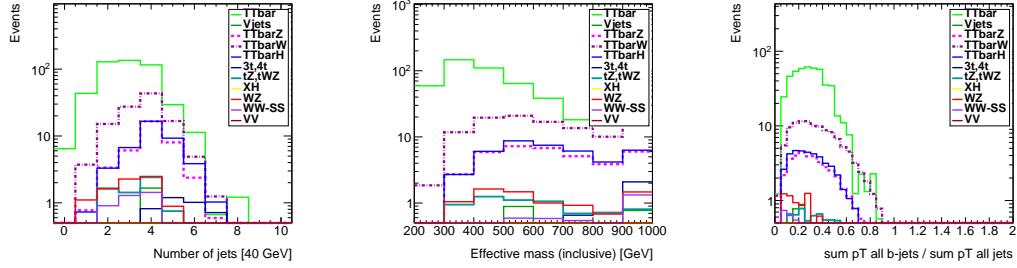


FIGURE 7.4: Number of jets with $p_T > 40$ GeV(left), m_{eff} (middle) and ratio between the sum of all b -jets p_T and the sum of all jets p_T (right) after lepton and jet selection of the $t\bar{t}W$ -VR (and no additional requirements). Signal regions are vetoed as detailed in Table 7.3. All MC samples are normalized to a luminosity of 36.1 fb^{-1} . Last bin includes overflow.

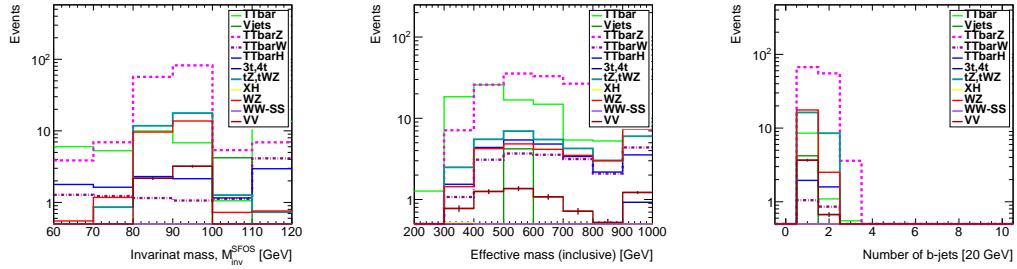


FIGURE 7.5: Invariant mass of the SFOS lepton pair (left) and m_{eff} (middle) after lepton and jet selection of the $t\bar{t}Z$ -VR (and no additional requirements). Number of b -jets with $p_T > 20$ GeV after the $t\bar{t}Z$ -VR selection. Signal regions are vetoed as detailed in Table 7.3. All MC samples are normalized to a luminosity of 36.1 fb^{-1} . Last bin includes overflow.

b -jets are required in the event. However, with such a cut the statistics will be highly reduced (up to a factor 2 lower as illustrated in Figure 7.5, right), so it is not pursued. The signal contamination is found to be around 5% for $\tilde{b}_1 \tilde{b}_1^*$ pair production.

7.2.3 Validation of irreducible background estimates

The observed yields, compared with the background predictions and uncertainties, can be seen in Table 7.4. There is good agreement between data and the estimated background in all the validation regions.

7.3 Reducible Backgrounds

The reducible backgrounds consist of the charge-flip background and the FNP lepton background. The charge flip background is obtained by re-weighting opposite-sign lepton pairs data with measured charge-flip probability and cross checked with the estimate obtained with the MC template method. Two data-driven methods are used to estimate the FNP lepton background, the matrix method and the MC template method, which are combined to obtain the final estimate. These methods were discussed in detail in Chapter 6. This section describes the application of these methods in the context of the SS/3L analysis and the validation of the reducible background estimates.

7.3.1 Charge-flip Background

The lepton charge mis-measurement commonly referred to as “charge flip”, is an experimental background strongly associated to analyses relying on same-sign leptons final states. In those events, the electric charge of one of the two leptons forming an opposite-sign (OS) pair, coming from an abundant SM process ($pp \rightarrow Z, t\bar{t}, W^+W^- \dots$), is mis-identified leading to a much rarer same-sign (SS)

pair event. In most cases, the source of such a mis-identification is the creation of additional close-by tracks $e^\pm \rightarrow e^\pm\gamma \rightarrow e^\pm e^\pm e^\mp$ via Bremsstrahlung of the original electron when interacting with the material of the inner tracker. If one of the secondary electron tracks is subsequently preferred to the original track in the reconstruction of the electron candidate, the charge assigned to the electron might be incorrect, leading to a charge-flip event. Errors on the track charge assignment itself may occur as well, but they are much rarer. In the case of muons, charge-flip is essentially negligible due to the much smaller interaction cross-section with matter, and the requirement of identical charges to be measured for the inner tracker and muon spectrometer tracks.

A purely data-driven method is used to estimate event yields for the electron charge-flip background. Assuming that the electron charge flip rates $\xi(\eta, p_T)$ are known, a simple way to predict these yields is to select events with pairs of opposite-sign leptons in data and assign them a weight :

$$w_{\text{flip}} = \xi_1(1 - \xi_2) + (1 - \xi_1)\xi_2 \quad (7.1)$$

where $\xi_{(i)} = 0$ for muons.

The advantages of this method are a good statistical precision since the charge flip rate is quite small, and the absence of dependency to the simulation and related uncertainties. Obviously, it requires a precise measurement of the rates, which is described in this section. An inconvenience of this approach is that the reconstructed momentum for charge-flipped electrons tends to be negatively biased

(too low by a few GeV), since such important Bremsstrahlung topologies represent only a very small fraction of the cases used to tune electron energy calibration. Simply re-weighting electrons from opposite-sign lepton pairs therefore does not predict correctly the charge-flip background shape for variables very sensitive to the electron momentum, for example the m_{ee} line-shape. However, the kinematic range and variables used in the analysis are not sensitive to this effect and can safely be neglected.

For the nominal (tight) estimate of the charge-flip background contributions, only events with exactly two OS signal electrons are considered. Corrections in the fake lepton estimate however require estimating as well charge-flip contributions for selections involving baseline electrons failing signal requirements; for that reason, the charge-flip (loose) rate is measured for these two categories of electrons.

Methodology

Charge-flip rates are measured in data relying on a clean $Z \rightarrow ee$ sample ($75 < m_{ee} < 100$ GeV), in which the rates can be determined from the relative proportions of OS and SS electron pairs. Figure 7.6 illustrates this event selection. The rates are measured as function of η and p_T , to follow their dependency to the distribution of material in the detector, the Bremsstrahlung emission rate, and the track curvature. Because of this binned measurements, and that the two electrons in a given pair generally have different kinematic properties, it has been found that the most efficient and less biased use of the available statistics is obtained by simultaneously extracting the rates in all bins via the maximization of the

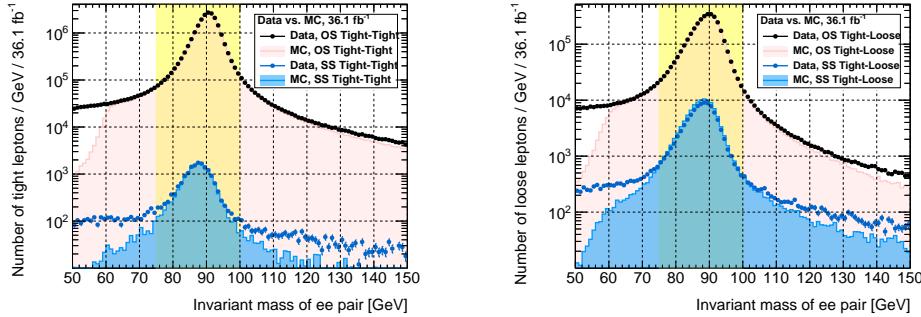


FIGURE 7.6: Invariant mass of opposite- and same-sign electron pairs, when both electrons satisfy signal requirements (left) or one of them fails them (right). Drell-Yan MC samples are not included, thus the drop in the MC distributions (light magenta filled area).

likelihood function describing the Poisson-expected yields of SS pairs :

$$L(\{N_{\varpi}^{\text{SS,obs}}\} | \{\xi(\eta, p_T)\}) = \prod_{\varpi} \mathcal{P}(N_{\varpi}^{\text{SS,obs}} | w_{\text{flip}}(\xi(\eta_1, p_{T,1}), \xi(\eta_2, p_{T,2})) \times N_{\varpi}^{\text{OS+SS,obs}}) \quad (7.2)$$

with $\varpi = (\eta_1, p_{T,1}, \eta_2, p_{T,2})$ indexing bins, where (arbitrarily) $p_{T,1} > p_{T,2}$; the expression of w_{flip} is given by (7.1). Statistical uncertainties on the extracted charge-flip rates are obtained (in a standard way) from the likelihood's numerically-computed Hessian matrix.

In the nominal charge-flip measurement, the two electrons are required to satisfy signal requirements. To measure charge-flip rates for baseline electrons failing signal (noted $\bar{\xi}$ below), pairs with only one signal electron are used ; this provides larger statistics than applying (7.2) to electrons pairs where both fail the signal cuts. However, the expression of the likelihood has to be adapted due

to the induced asymmetry between the two electrons forming the pair :

$$\begin{aligned} L(\{N_{\varpi}^{\text{SS,obs}}\} | \{\xi(\eta_1, p_{\text{T},1})\}, \{\bar{\xi}(\eta_2, p_{\text{T},2})\}) = \\ \prod_{\varpi} \mathcal{P}(N_{\varpi}^{\text{SS,obs}} | w_{\text{flip}}(\xi(\eta_1, p_{\text{T},1}), \bar{\xi}(\eta_2, p_{\text{T},2})) \times N_{\varpi}^{\text{OS+SS,obs}}) \end{aligned} \quad (7.3)$$

where this time $(\eta_1, p_{\text{T},1})$ corresponds to the signal electron. Using the same η and p_{T} binning for both measurements, the number of free variables in the maximization of (7.3) – as well as the number of terms in the product forming L – is twice larger than for the nominal case (7.2). In fact, a by-product of the maximization of (7.3) is another determination of the charge-flip rates for signal electrons, although with a more limited precision than obtained in the nominal measurement (7.2).

Background subtraction is performed through a simple linear extrapolation of the invariant mass distribution sidebands ; it matters mostly for low p_{T} in the nominal measurement, and for the additional measurement with baseline electrons failing signal requirements, where the level of background is larger.

Measured rates

The charge-flip rates measured in data and MC are shown on Figure 7.7. In data, the nominal rates (Fig. 7.7a) go up to $\sim 0.1\%$ in the barrel region ($|\eta| < 1.37$), while it increases up to $\sim 0.2\%$ in the end-cap region ($|\eta| > 1.37$). For baseline electrons failing signal requirements (Fig. 7.7c), the rates are in general greater than the nominal ones in every bin, as expected. The charge-flip rates for these electrons go up to $\sim 0.5\%$ in the barrel region and up 1% in the end-cap region. Compared to the rates used in the previous version of the analysis [79], the

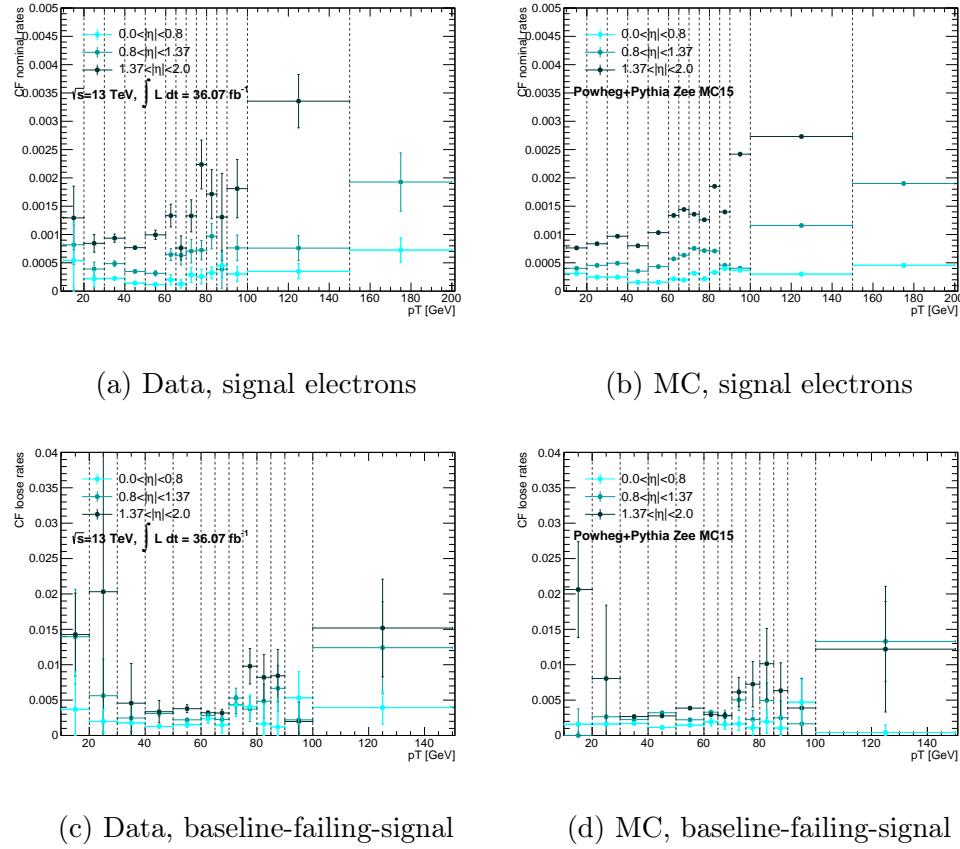


FIGURE 7.7: Charge-flip rate as measured in data (left) and MC (right). Only the statistical uncertainty is displayed. The last p_T bin is inclusive.

central values are much lower now. After suppressing the charge flip events with the charge-flip electron BDT classifier described in Section 5.4.3, the charge flip rates are strongly reduced for both signal and baseline-failing-signal electrons (up to a factor 20 in some bins). Figure 7.8 illustrates the charge flip background reduction in a loose selection. Below 30 GeV the statistics is very low for the loose measurement ; however, these results are used only to measure the electron fake rate and, as illustrated in Figure 7.14, in this p_T interval the charge flip background is negligible.

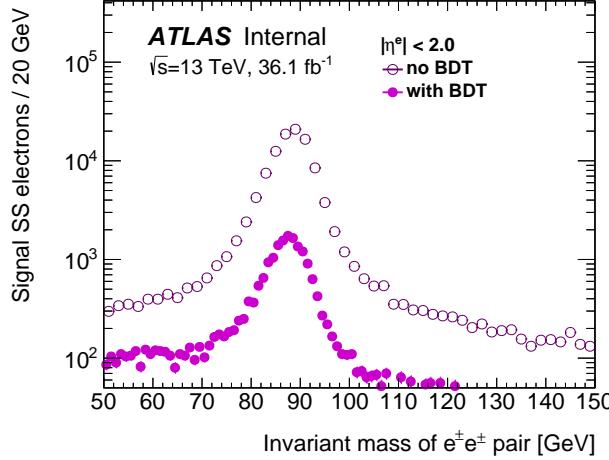


FIGURE 7.8: Invariant mass of the signal $e^{\pm}e^{\pm}$ pair distribution with (full markers) and without (open markers) charge-flip electron BDT selection applied.

The charge-flip rates in MC (Figs. 7.7b, 7.7d) are obtained by applying the same methodology as in data. Generally, the rates are not very far from data, validating the use of MC to predict charge-flip background in several of the optimization studies presented in this document. In addition, a closure test is performed on $t\bar{t}$ MC, checking that weighted OS events can reproduce the distribution of SS charge-flip events (identified by truth-matching). A good overall agreement is found, largely within the assigned uncertainties as shown in Figure 7.9.

Systematic uncertainties

The main uncertainties on the measured charge-flip rates come from the presence of background and the way it is estimated. To assess them, variations of the selection and background estimation are considered :

- 1) $75 < m_{ee} < 100$ GeV, no background subtraction ;

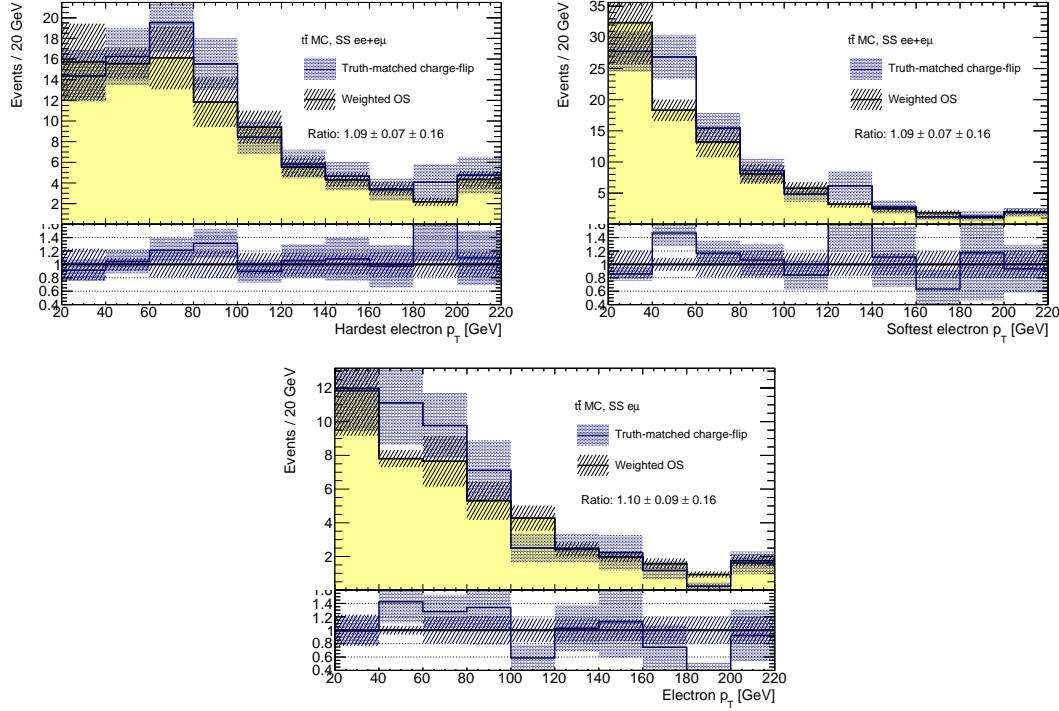


FIGURE 7.9: Closure test for the charge-flip background prediction, for simulated $t\bar{t}$ events using charge-flip rates measured in $Z \rightarrow ee$ MC (with the systematic uncertainties from the data measurements, though). Events are selected in the $e\mu$ and ee channels, using signal leptons only, and charge-flipped electrons are identified by truth-matching.

- 2) $75 < m_{ee} < 100$ GeV, sidebands of 20 GeV ;
- 3) $75 < m_{ee} < 100$ GeV, sidebands of 25 GeV(nominal measurement) ;
- 4) $75 < m_{ee} < 100$ GeV, sidebands of 30 GeV ;
- 5) $80 < m_{ee} < 100$ GeV, sidebands of 20 GeV.

The effect of applying the background subtraction itself is evaluated by comparing configurations 1 and 3. The impact of the width of the m_{ee} chosen for the measurement is by comparing configurations 3 and 5, while the sideband width

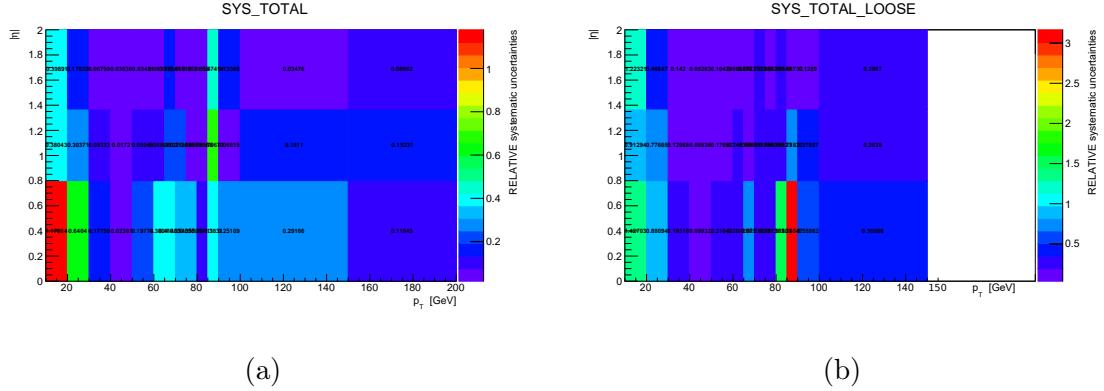


FIGURE 7.10: Total systematic uncertainties on the charge-flip rates for electrons satisfying the signal requirements (7.10a), and for baseline electrons failing the signal requirements (7.10b).

effects are evaluated by comparing configuration 3 and 2, or 3 and 4. The largest deviation in each bin is taken as the systematic uncertainty on the charge-flip rate.

The resulting systematic uncertainties on the charge-flip rates are presented in Fig. 7.10.

For the signal electrons charge-flip rates the systematic uncertainties vary in general between 2% and 20% (increasing up to $> 50\%$ in the region with $p_T < 10$ GeV), whereas for baseline-failing-signal electrons they vary between 3% and 30% (increasing up to $> 50\%$ in the region with $p_T < 10$ GeV). Part of these large values, at low p_T and in the [80,90] GeV p_T interval, can be explained by large statistical fluctuations between the different configurations.

7.3.2 Matrix Method

The matrix method relates the number of events containing prompt or FNP leptons to the number of observed events with tight or loose-not-tight leptons using the probability for loose prompt or FNP leptons to satisfy the tight criteria. The formalism for this method has been discussed in Section 6.4. The next sections will concentrate on the measurement of the two input variables needed for the matrix method : the probability for loose FNP leptons to satisfy the tight selection criteria (ζ) and the probability for loose prompt leptons to satisfy the tight selection criteria (ε).

Baseline-to-signal efficiency for fake muons

Baseline-to-signal efficiency for fake leptons (further called “fake rate”, (ζ)) is measured in a sample enriched in fake leptons from $t\bar{t}$ processes. The MC simulations indicate that this background has the largest contribution to FNP lepton background in the signal regions, even those with b -jet vetoes, due to the requirements on jet multiplicity and E_T^{miss} . The events used for the measurements require exactly two same-sign muons (and no extra baseline lepton), at least one b -jet, and at least 3 jets that were acquired by di-muon triggers. One of the muons in the event (referred to as “tag”) is required to satisfy signal requirements, verify $p_T > 25$ GeV, and trigger the event recording. The measurement may then be performed on the other lepton (“probe”), likely to be the fake lepton of the pair.

Figure 7.11a shows the number of signal muon probes available after this

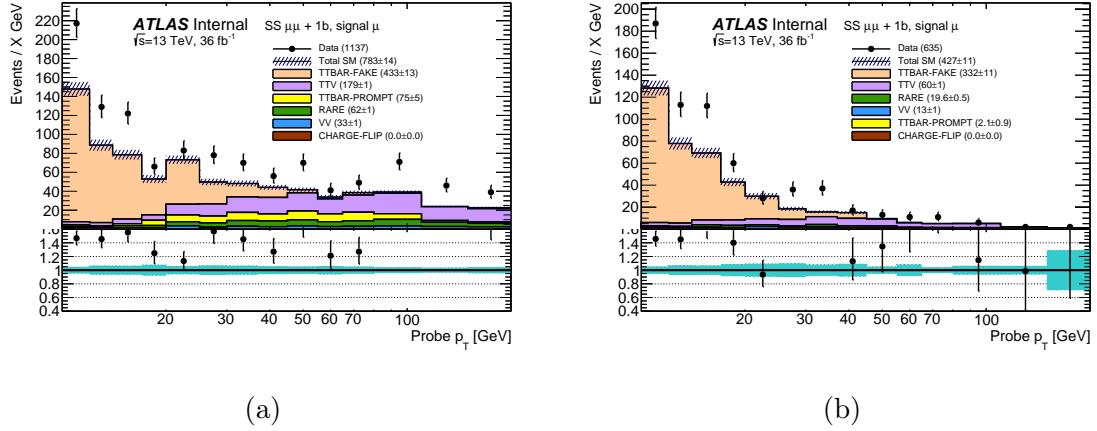


FIGURE 7.11: Signal probe muon p_T distribution in data and MC, after pre-selection (left) or further tightening of the tag muon requirements (right). The yellow area indicates $t\bar{t}$ events in which the tag muon is fake and the probe real, leading to a measurement bias.

pre-selection. It is clear that at this stage, measurements above 25 GeV would be very affected by the important fraction of events in which the tag muon is fake and the probe muon is real. To overcome this issue, two alternatives are considered :

- tighten the p_T and isolation requirements of the tag muon beyond the “signal” requirements, to reduce its probability of being a fake muon
- use an identical selection for tag and probe muons, and require them to be in the same (p_T, η) bin for the measurement ; after subtraction of estimated contributions from processes with two prompt muons, all events have one real and one fake muon, and the symmetry in the muon selection can be

taken advantage of to obtain an unbiased measurement of the fake rate :

$$\zeta = \frac{\varepsilon n_2}{\varepsilon n_1 + (2\varepsilon - 1)n_2}$$

with n_1, n_2 the number of events with 1 or 2 signal muons, and ε the efficiency for prompt muons.

This method is limited to measurements in inclusive or wide bins. It also can't be used at too low p_T , due to contributions from processes with two fake muons (e.g. from $B\bar{B}$ meson production).

Comparisons made with $t\bar{t}$ MC indicated that when using a very tight isolation requirement on the tag muon ($\max(E_T^{\text{topo, cone 40}}, p_T^{\text{cone 40}}) < 0.02 \times p_T$), the level of bias is always largely inferior to the statistical uncertainty in the measurement, which itself is smaller than for the other two methods.

Figure 7.11b shows the number of signal muon probes when applying those reinforced isolation criteria to the tag muon, as well as requiring $p_T^{\text{tag}} > \max(40, p_T^{\text{probe}} + 10)$ GeV. As expected, the number of pairs with a fake tag muons is down to a minor level, at least according to the simulation.

Muon fake rates as predicted by the simulation ($t\bar{t}$, inclusive selection of leptons via truth-matching) are shown on Fig. 7.12 as function of p_T and $|\eta|$. One can expect a moderate dependency of the fake rates to the transverse momentum, with the strongest evolution at low p_T and a slight increase toward higher p_T . The fake rates are also essentially independent of the pseudorapidity, except at the edge ($|\eta| > 2.3$) where there is a strongly pronounced increase of the rates. This motivates measurements in data as function of p_T in two $|\eta|$ bins.

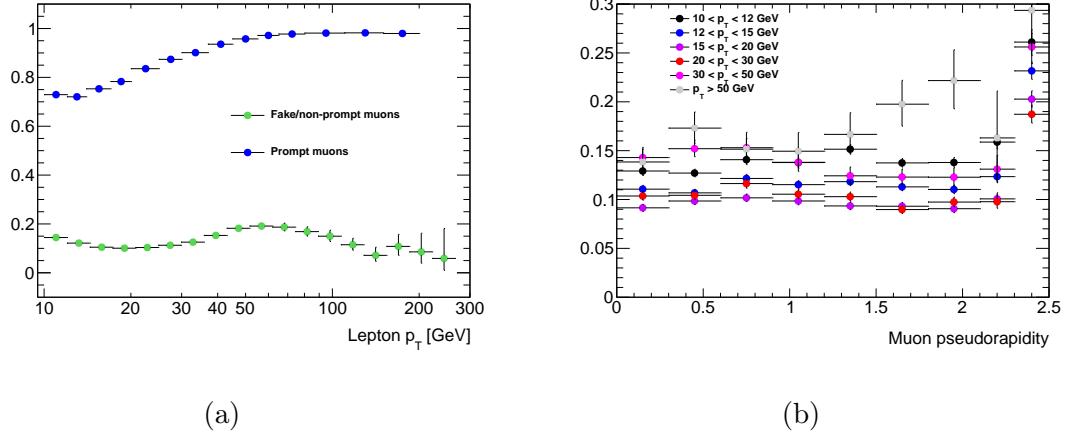


FIGURE 7.12: Muon fake rates in $t\bar{t}$ MC with an inclusive selection, as function of p_T (left, green markers) or $|\eta|$ in different momentum ranges (right).

Observations in data seem to indicate that the rejection of fake tag muons by the reinforced isolation criteria is quite less important than in the simulation, or that the amount of fake muons at high p_T is larger than in the simulation, or both. This leads to an unknown level of bias in measurements performed with the straightforward tag-and-probe selection at high p_T . For that reason, the final rates measured in data are provided by the tag-and-probe method below 25 GeV, and by the symmetric selection for $p_T > 25$ GeV. The former are obtained with

$$\zeta = \frac{n_{\text{signal}}^{\text{data}} - n_{\text{signal}}^{\text{MC}}}{n_{\text{baseline}}^{\text{data}} - n_{\text{baseline}}^{\text{MC}}} \quad (7.4)$$

with $\Delta\zeta_{\text{stat}} = \frac{\sqrt{(1-2\zeta)n_{\text{signal}}^{\text{data}} + \zeta^2 n_{\text{baseline}}^{\text{data}}}}{n_{\text{baseline}}^{\text{data}} - n_{\text{baseline}}^{\text{MC}}}$

while the latter are obtained with :

$$\zeta = \frac{\varepsilon(n_{\text{both signal}}^{\text{data}} - n_{\text{both signal}}^{\text{MC}})}{\varepsilon(n_{\text{only 1 signal}}^{\text{data}} - n_{\text{only 1 signal}}^{\text{MC}}) + (2\varepsilon - 1)(n_{\text{both signal}}^{\text{data}} - n_{\text{both signal}}^{\text{MC}})} \quad (7.5)$$

with $\Delta\zeta_{\text{stat}} = \frac{\zeta}{n_{\text{both signal}}^{\text{data}} - n_{\text{both signal}}^{\text{MC}}} \sqrt{\zeta^2 n_{\text{only 1 signal}} + \left(1 - \frac{2\varepsilon - 1}{\varepsilon}\zeta\right)^2 n_{\text{both signal}}}$

the efficiency for prompt muons ε is assigned values compatible with section 7.3.2.

The measured rates are presented in Table 7.5. The central values are shown together with the associated statistical uncertainty, as well as the propagation of the uncertainty on the subtracted backgrounds normalization, which is taken as a global $\Delta B/B = 20\%$. The rates are of the order of 10% up to 30 GeV, beyond which they increase. Overall these values are not very different from those predicted by the simulation.

Some of the validation and signal regions require events with 2 or more b -tagged jets, which reduces the fraction of non-prompt muons coming from B meson decays. Figure 7.13 illustrates how this impacts on the fake rates. Given the good agreement between data and simulation for the measured values, A correction is applied to the measured rates for events with ≥ 2 b -jets, taken directly from simulated $t\bar{t}$ events. This correction factor varies between 1 and 2 with p_T , and the whole size of the correction is assigned as an additional systematic uncertainty (see Table 7.6).

Systematic uncertainties

To cover potential differences in the fake rates between the measurement regions and the signal regions, that could be due to different origins or kinematic properties of the fake leptons, uncertainties are set based on the extent of those differences

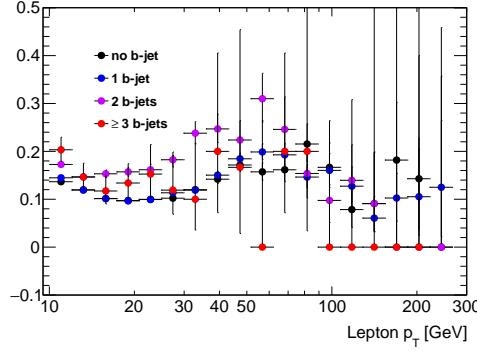


FIGURE 7.13: Muon fake rates in $t\bar{t}$ MC with an inclusive selection, as function of p_T and split according to the number of b -tagged jets in the event.

predicted by the simulation. The largest effect is the decrease of the fake rates with H_T (especially for high- p_T muons), which likely correlates to a harder jet at the origin of the non-prompt muon, hence a reduced likelihood to satisfy isolation requirements. Table 7.6 summarizes the additional systematic uncertainties applied to the muon fake rates. They vary from 30% at low p_T , to up to 85% for $p_T > 40$ GeV ; in that range, the uncertainties are made H_T -dependent.

As already shown, Fig. 7.13 shows the variation of the fake rate in $t\bar{t}$ MC as function of the number of b -tagged jets in the event. Unsurprisingly, the rates are very similar for $0b$ and $\geq 1b$ final states, justifying the use of the fake rates measured in this section (i.e. in a $\geq 1b$ region) to predict fake muon background in all signal regions.

Baseline-to-signal efficiency for fake electrons

Electron fake rates are measured with a similar methodology, but the $e^\pm e^\pm$ channel is unusable due to the presence of a large charge-flip background. This

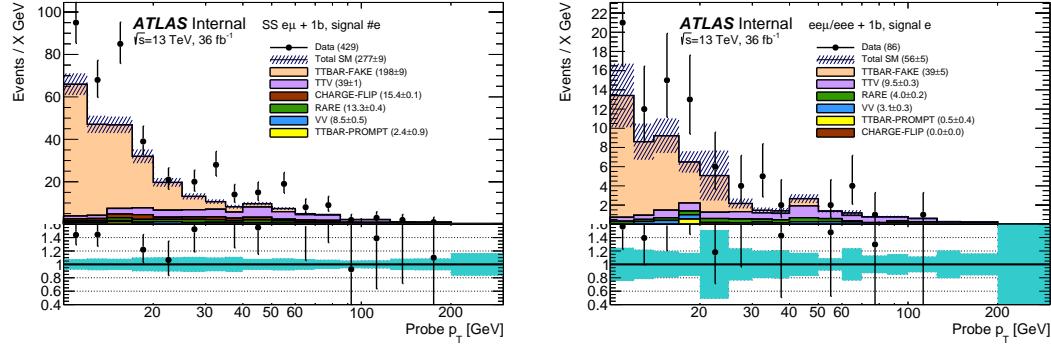


FIGURE 7.14: Signal probe electron p_T distribution in data and MC, for $e^\pm\mu^\pm$ pairs (left, with probe electrons satisfying CFT cut and a reinforced tag muon selection) or $\ell^\pm e^\mp e^\mp$ pairs (right, with reinforced tag electron selection), as described in section 7.3.2. The yellow area indicates $t\bar{t}$ events in which the tag lepton is fake and the probe electron real, leading to a measurement bias.

is overcome by working with $e^\pm\mu^\pm$ pairs instead (with a tag muon), but mixing leptons of different flavours brings additional complications (for example, the unbiased measurement cannot be employed to measure muon fake rates at higher p_T , as there is no symmetry between the leptons). To improve confidence, measurements are performed in four different ways, which complement each other :

- straightforward tag-and-probe with $e\mu$ pairs, with the same tag muon selection as in the previous section.
- same selection, but subtracting from the numerator the number of pairs with one fake tag muon and one prompt probe electron, itself estimated from the number of observed $e\mu$ events with a muon failing signal requirements, scaled by an efficiency correction factor $e\mu/\mu\mu$ taken from $t\bar{t}$ MC (only for pairs with one fake muon). This only works if the two muons satisfy the

same kinematic requirements, therefore can be used only for measurements in wide or inclusive bins.

- selecting $\ell^\mp e^\pm e^\pm + \geq 1b$ events, with a Z veto on SFOS pairs. This selection entirely suppresses contributions from charge-flip, or events with fake muons. One of the electron, with standard signal requirements, is required to satisfy the same reinforced p_T and isolation requirements as for the muon measurement, and the measurement can be performed on the other electron.
- same selection, using the symmetry between the two same-sign electrons to measure the rates in an unbiased way, similarly to the muon case.

Events are acquired with the combination of single-muon (as in previous section) and $e\mu$ triggers.

Figure 7.14 shows the number of signal probe electrons selected in the $e\mu$ and ℓee channels. There are quite less events selected in the trilepton channel. Figure 7.15 shows the electron fake rate as a function of p_T or η in $t\bar{t}$ MC. The variations of the rates as function of the pseudorapidity are not very large, therefore measurements are only performed as a function of p_T . The low p_T range is dominated by non-prompt electrons from heavy flavour decays, while beyond 30 GeV, electron fakes mostly come from conversions of photons produced inside jets, such as $\pi^0 \rightarrow \gamma\gamma$ decays.

Based on the estimated levels of bias, and achievable statistical precision, of the different methods, the electron fake rate is measured with the tag-and-probe $e\mu$ selection up to 30 GeV, and by combining “unbiased” evaluations in both $e\mu$

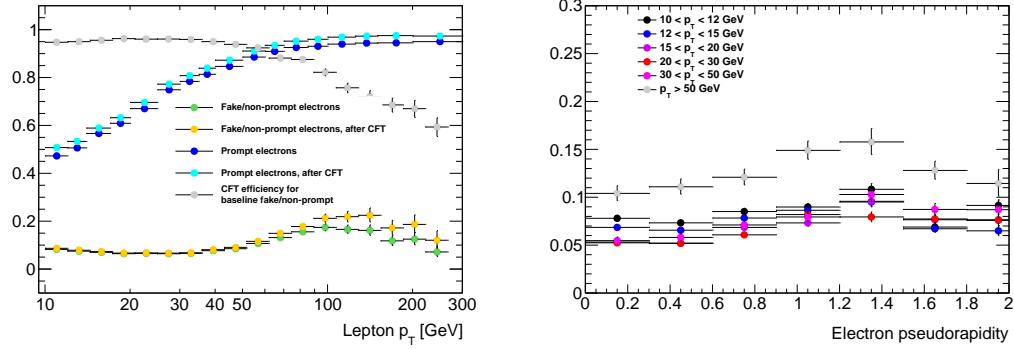


FIGURE 7.15: Electron fake rates in $t\bar{t}$ MC with an inclusive selection, as function of p_T (left, yellow/green markers = with/without CFT cut applied) or $|\eta|$ in different momentum ranges (right).

and ℓee channels beyond. The measured rates are presented in Table 7.7, together with the associated statistical and background-subtraction uncertainties. The rates are of the order of 10% up to 30 GeV, beyond which they increase to up to 25%.

Unlike muons, MC-based correction factors are not applied for final states with ≥ 2 b -tagged jets. This is because there is less good agreement between the measured rates and the simulation ; in particular the former take larger values in the medium- p_T range.

Systematic uncertainties

Similarly to the muon case, systematic uncertainties are assigned to cover for difference in the rates in the measurement regions and in the signal regions that would be due to different sources of fake leptons, or different kinematic properties of these sources. Unlike muons, there is much less of a dependency to H_T . The

dominant source of potential differences is therefore the origin of the fake electron (see Fig. 7.16) ; for $p_T < 20$ GeV, non-prompt electrons from heavy-flavor hadron decays dominate, which is confirmed by the good agreement between MC fake rates and those measured in data. In that range, an uncertainty of 30% is assigned to the fake rates (inflated to 50% for final states with $\geq 2b$ -tagged jets). The rates measured in data are larger than those predicted by the simulation, and would for example be consistent with a larger amount of electrons from photon conversions than predicted. In that range, an uncertainty of 50% is assigned to cover any arbitrary variation of the relative contributions of each source.

Finally, Figure 7.17 shows the variation of the fake rate in $t\bar{t}$ MC as function of the number of b -tagged jets in the event. As expected, the rates are very similar for $0b$ and $\geq 1b$ final states, justifying the use of the fake rates measured in this section (i.e. in a $\geq 1b$ region) to predict fake electron background in all signal regions.

Baseline-to-signal efficiency for real leptons

Baseline-to-signal efficiency for real leptons is measured in a high purity data sample of opposite-sign same-flavor leptons with the standard Z tag-and-probe method. Events are selected by a single lepton trigger. The tag lepton, required to have triggered the event recording, also satisfies signal requirements and verifies $p_T > 25$ GeV. The probe lepton used for the efficiency measurement satisfies baseline requirements. All possible tag-and-probe combinations are considered in an event (including permutation of the tag and probe leptons), as long as the

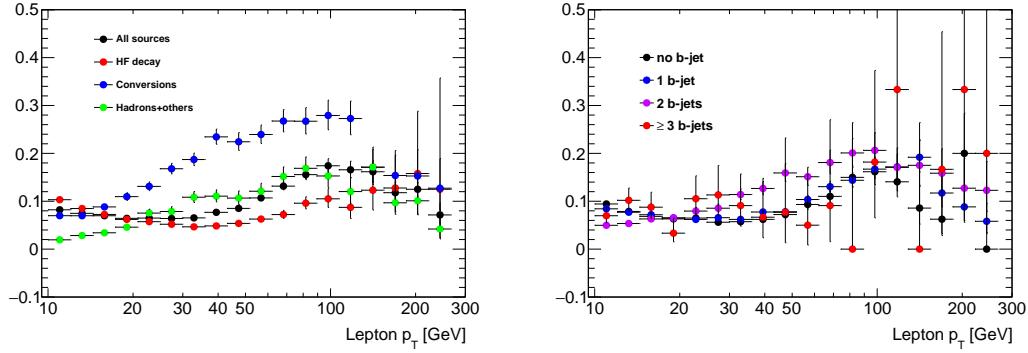


FIGURE 7.16: Electron fake rates in $t\bar{t}$ MC with an inclusive selection, as function of p_T and split according to the source of the fake electron (left). The relative contributions of each source (for signal electrons) are indicated on the right-hand-side.

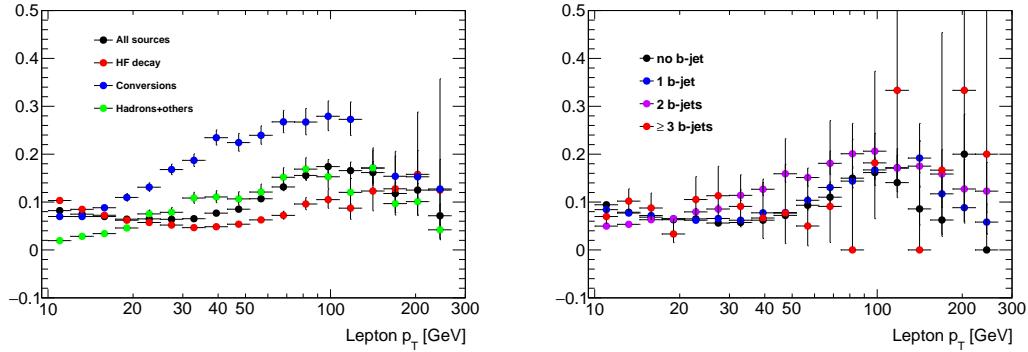


FIGURE 7.17: Electron fake rates in $t\bar{t}$ MC with an inclusive selection, as function of p_T and split according to the number of b -tagged jets in the event.

invariant mass of the pair is comprised between 80 and 100 GeV. Figure 7.18 illustrates this event selection.

A non-negligible background contamination in the electron channel affects measurements below $p_T = 20$ GeV. This contamination is taken into account in the measurement using a background template method inspired by the method used

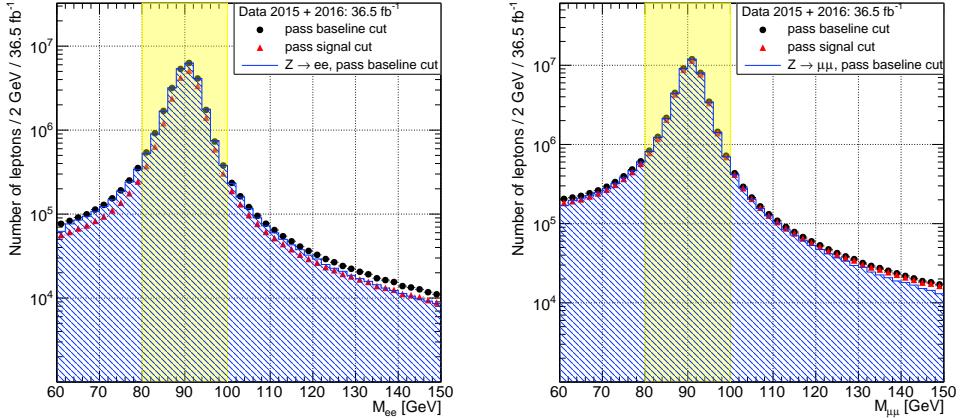


FIGURE 7.18: Invariant mass of opposite-sign same-flavor electrons (left) and muons (right), after the tag selection, where the probe satisfies the baseline requirements or the signal requirements.

to measure reconstruction, identification, and isolation efficiencies documented in [80]. This template is built from the tag-and-probe invariant mass distribution for baseline-level probe electrons that fail both tight identification and isolation requirements, smoothed by assuming an exponential shape whose parameters are determined by a fit in the interval $60 < m_{ee} < 120$ GeV excluding the $80 < m_{ee} < 100$ GeV region. The background template is then normalized to the main tag-and-probe distribution in the background-dominated tail $120 < m_{ee} < 150$ GeV. The estimated level of background goes up to 4%, reached for probe electrons with $p_T < 15$ GeV and $|\eta| < 0.8$.

The efficiency is measured as a function of p_T and η , and the results are presented in Fig 7.19 for electrons and muons. The background subtraction is applied on the electron channel only. The following systematic uncertainties are

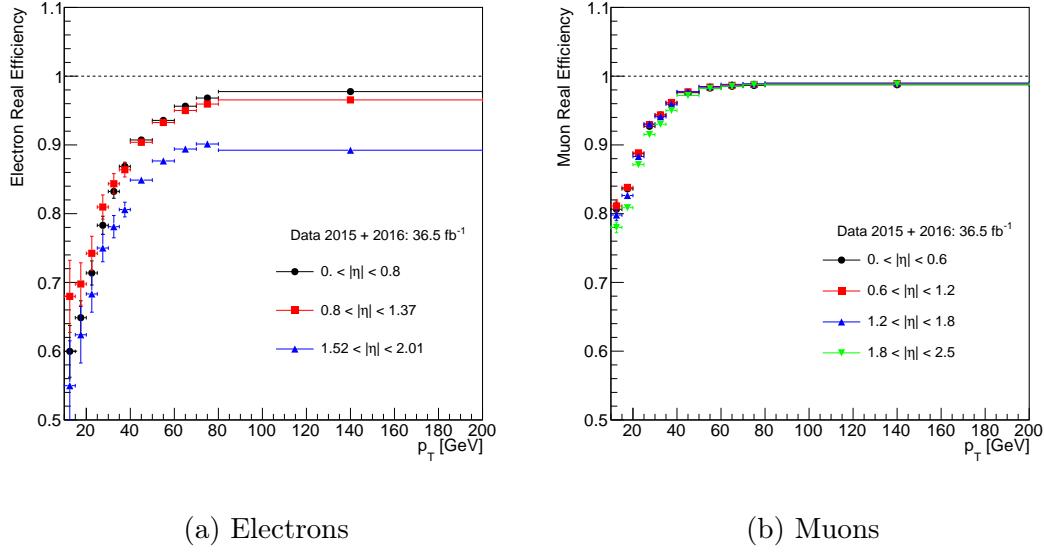


FIGURE 7.19: Baseline-to-signal efficiencies as a function of p_T and $|\eta|$ for real electrons (left) and muons (right), measured in 2015+2016 data. The $|\eta|$ binning used in the electron case corresponds to the geometry of the electromagnetic calorimeter. For muons a homogeneous $|\eta|$ binning is considered. The error bars corresponds to the quadratic sum of the statistical and tag-and-probe measurement systematic uncertainties.

assigned to the measured efficiencies :

- Background contamination : 27 variations of the tag-and-probe method are considered to assess the electron measurement systematics. Three m_{ee} windows and 9 variations of the background subtraction methods are considered. The largest contribution to the systematics arises from the m_{ee} window variation. This is expected as the proportion of electrons affected by Bremsstrahlung depends on m_{ee} . The resulting relative systematics vary from $6\% \sim 12\%$ in the $10 < p_T < 15$ GeV region, 3% to 6% in

the $15 < p_T < 20$ GeV region, 1% to 3% in the $20 < p_T < 40$ GeV region, and less than 1% for $p_T > 40$ GeV. The systematic uncertainties associated to the muon efficiencies measurement vary from 1% to 1.3% in the $10 < p_T < 15$ GeV region and less than 1% for $p_T > 15$ GeV.

- Trigger : a systematic uncertainty accounting for a potential bias at trigger level is considered and it varies between 0 and 4%, depending on the p_T range.
- Extrapolation to busy environments : efficiencies are typically lower in such environments due to the proximity of jets and leptons ; an uncertainty is assigned by comparing efficiencies in simulated $Z \rightarrow \ell\ell$ and $\tilde{g} \rightarrow t\bar{t}\tilde{\chi}_1^0$ events, for $\Delta m(\tilde{g}, \tilde{\chi}_1^0) > 1$ TeV which represents an extreme case of final states with highly boosted top quarks. The uncertainty, taken as the difference in efficiencies, is parametrized as a function of p_T and ΔR (the angular distance between the lepton and the closest jet).

The resulting systematic uncertainties are summarized in Table 7.8 and Table 7.9.

7.3.3 MC Template Method

As discussed in Section 6.3.1, the MC template method is a simulation-based method that provides an alternative estimate of the reducible backgrounds affecting the analysis. It relies on the correct modelling of FNP leptons and charge-flipped

electron kinematics in $t\bar{t}$ and $V+jets$. FNP leptons are classified in four categories, namely electrons and muons coming from b and light-quark jets. Normalisation factors for each of the five sources (including photon conversions) are computed to match the observed data in dedicated control regions. The fifth category is events with prompt electrons that have a charge mis-measurement (charge flip, EL HF, MU HF, EL LF, MU LF). Six non-overlapping control regions are defined by the presence of b -jets and by the flavors of the same sign lepton pair in the event :

- CR0b : events without b -jets in ee , $e\mu$, and $\mu\mu$ channels.
- CR1b : events with at least one b -jet in ee , $e\mu$, and $\mu\mu$ channels.

All the selected events contain two or more same-sign signal leptons and $E_T^{\text{miss}} > 40$ GeV and 2 or more jets. Events satisfying the signal regions requirements are excluded from the control regions. The purpose of the E_T^{miss} requirement is to remove multi-jet events that have two or more FNP leptons and tend to have low E_T^{miss} . The six distributions are chosen for variables that provide the best separation between processes with prompt leptons and processes with fake leptons and charge flip and are shown before and after the fit in Figures 6.2-6.4 and Figures 6.3-6.5, respectively. The multipliers obtained after the minimization of the negative log likelihood were given in Tables 6.1 and 6.2. The tables represent the correction factors obtained from the fit upon using two different parton showers, Pythia and Sherpa for the processes that lead to non-prompt leptons and charge flips. The goal of varying the parton shower is to access the dependence of the fake and charge flip estimates on the choice of the parton shower.

The MC template method is validated by looking at the agreement between observed data and prediction as shown in Figure 7.23 and Figure 7.23. In the MC template method, the systematic uncertainty is obtained by changing the generator from POWHEG-BOX to SHERPA and propagating uncertainties from the control region fit to the global normalization scale factors applied to the MC samples. The uncertainties in these scale factors are in the range 75–80%, depending on the SRs. In practice, only $t\bar{t}$ contributes to the SRs and the final yields with systematic uncertainties from fit uncertainty, theory uncertainties on $t\bar{t}$, comparison of different showers (Pythia 6 and Sherpa) is shown in Table 7.10. This table also shows a global correction factor derived by taking the ratio of the weighted $t\bar{t}$ to raw MC $t\bar{t}$ with a global uncertainty that includes all systematic uncertainties used to obtain the final estimate.

7.3.4 Reducible Background Validation

The reducible backgrounds estimated with the methods described in the previous sections are validated by comparing observed data to predicted background after various kinematic requirements. The next sections will validate the MC template method used to estimate the charge flip and the FNP lepton background, and the matrix method used to estimate the FNP lepton background.

Data/MC comparisons

The overall level of agreement obtained by the matrix method can be seen in different channels in Fig 7.20. The distributions of several variables are shown on

Fig 7.21 for an inclusive same-sign leptons selection after various requirements on the number of jets and b -jets. These distributions illustrate that the data are described by the prediction within uncertainties. The apparent disagreement for m_{eff} above 1 TeV in Figure 7.21d is covered by the large theory uncertainty for the diboson background, which is not shown but amounts to about 30% for m_{eff} above 1 TeV. Similar distributions for the other main variables can be found in Appendix B. To avoid the presence of signal in the tails of these distributions, events belonging to any of the signal regions are vetoed both in data and for the predicted backgrounds.

Figure 7.22 shows a comparison between the estimates of the MC template method and the matrix method in a loose selection. Other $E_{\text{T}}^{\text{miss}}$ distributions with events satisfying the signal region requirements except the $E_{\text{T}}^{\text{miss}}$ cut are shown in Figure 7.23 comparing the two methods.

Reducible background estimates

The expected yield for processes with FNP leptons and charge-flip electrons, estimated with the matrix method, likelihood method (charge-flip), and the MC template method are presented in Table 7.11 and Table 7.12 for the signal regions and Table 7.13 for the validation regions. Since the predictions from the MC template and matrix methods in the signal and validation regions are consistent with each other, the final numbers retained for the FNP lepton background estimate (also shown in the tables) are taken as the weighted-average of the predictions from the matrix method and the MC template; the weights are based

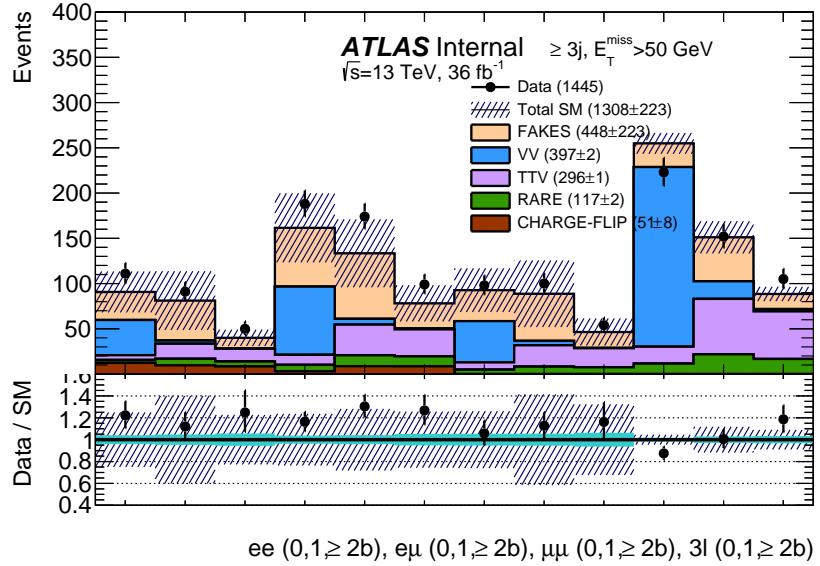


FIGURE 7.20: Summarized level of agreement between observed data (2015+2016, 36.5 fb^{-1}) and expected SM+detector backgrounds for events with ≥ 2 same-sign leptons ($p_T > 20\text{GeV}$), $E_T^{\text{miss}} > 50\text{GeV}$ and ≥ 3 jets ($p_T > 40\text{GeV}$), split as function of the lepton flavours and the number of b -tagged jets. Uncertainties include statistical sources, as well as systematic uncertainties for the data-driven backgrounds ; for illustration, statistical uncertainties alone are shown in the light-coloured error bands in the ratio plots. Events belonging to any of the signal regions are rejected, both in data and MC.

on the statistical component, and the systematic uncertainties are propagated assuming conservatively a full correlation between the two methods (although they are in fact largely independent!). The central value and statistical/systematic uncertainties are therefore :

$$(w\zeta_1 + (1-w)\zeta_2) \pm \sqrt{w^2 \left(\Delta\zeta_1^{(\text{stat})} \right)^2 + (1-w)^2 \left(\Delta\zeta_2^{(\text{stat})} \right)^2} \quad (7.6)$$

$$\pm \left(w\Delta\zeta_1^{(\text{syst})} + (1-w)\Delta\zeta_2^{(\text{syst})} \right)$$

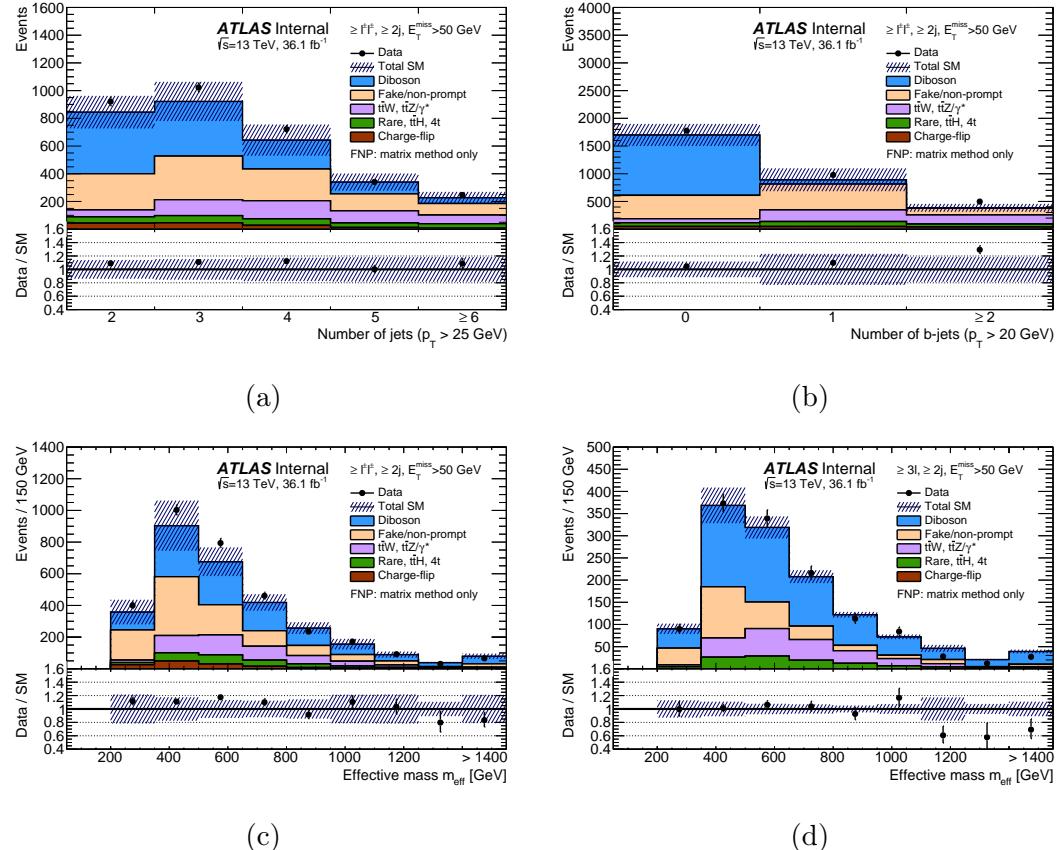


FIGURE 7.21: Distributions of (a) the number of jets, (b) the number of b -tagged jets and (c), (d) the effective mass. The distributions are made after requiring at least two jets ($p_T > 40\text{GeV}$) and $E_T^{\text{miss}} > 50\text{GeV}$, as well as at least two same-sign leptons ((a), (b), (c)) or three leptons (d). The uncertainty bands include the statistical uncertainties for the background prediction as well as the systematic uncertainties for fake- or non-prompt-lepton backgrounds (using the matrix method) and charge-flip electrons. Not included are theoretical uncertainties in the irreducible background contributions. The rare category is defined in the text.

$$\text{with } w = \frac{\left(\Delta\zeta_2^{(\text{stat})}\right)^2}{\left(\Delta\zeta_1^{(\text{stat})}\right)^2 + \left(\Delta\zeta_2^{(\text{stat})}\right)^2}$$

When the estimated value is too small(below 0.15), the expected yield is set

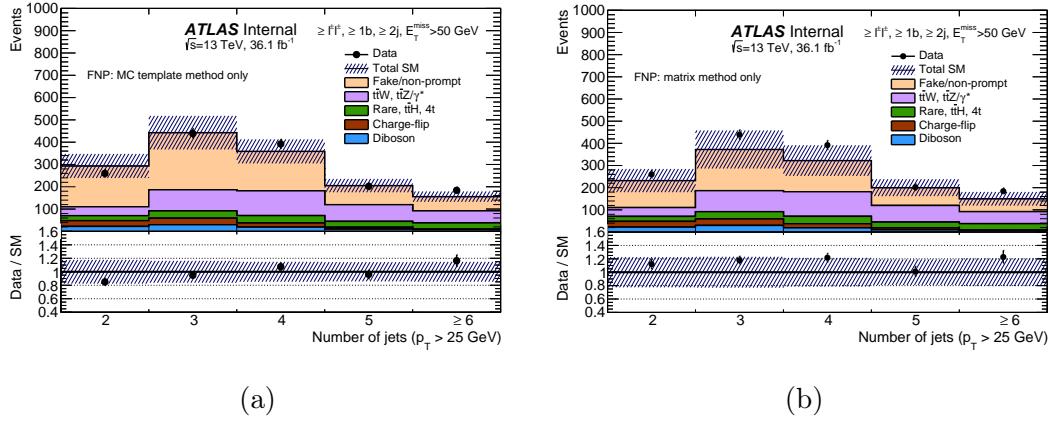


FIGURE 7.22: Distributions of the number of jets after requiring at least two jets ($p_T > 40\text{GeV}$) and $E_T^{\text{miss}} > 50\text{GeV}$, as well as at least two same-sign leptons. The fake or non-prompt leptons backgrounds are estimated alternatively with the MC template method (7.22b) or the matrix method (7.22a). The uncertainty band includes the statistical uncertainties for the background prediction as well as the full systematic uncertainties for fake or non-prompt leptons backgrounds or charge-flip electrons. The rare category is defined in the text. In both figures, the last bin contains the overflow.

to 0.15 ± 0.15 , to cover for possibilities of an under-fluctuation of the number of baseline-not-signal leptons when applying the matrix method, as well as lack of statistics in the MC samples for the other method.

The charge flip background is not combined between the MC template method and the likelihood method due to the very large uncertainty in the MC template estimate. The OS data has a much larger number of events which makes a precise prediction of this background. The MC template result is used as a cross check.

7.4 Systematic uncertainties

The systematic uncertainties related to the estimated background from same-sign prompt leptons arise from the experimental uncertainties as well as theoretical modelling and theoretical cross-section uncertainties. The statistical uncertainty of the simulated event samples is also taken into account.

7.4.1 Theoretical Uncertainties

The cross-sections used to normalize the MC samples are varied according to the uncertainty in the cross-section calculation, which is 13% for $t\bar{t}W$, 12% for $t\bar{t}Z$ production [48], 6% for diboson production [50], 8% for $t\bar{t}H$ [48] and 30% for $4t$ [22]. Additional uncertainties are assigned to some of these backgrounds to account for the theoretical modelling of the kinematic distributions in the MC simulation.

Associate $t\bar{t} + W/Z$ production

The theoretical uncertainties on the ttW and $ttZ^{(*)}$ processes are evaluated by several variations added in quadrature :

- Normalization and factorization scales varied independently up and down by a factor of two from the central scale $\mu_0 = H_T/2$ as detailed in Ref. [58].

The largest deviation with respect to the nominal is used as the symmetric uncertainty.

- Variation of the PDF used. The standard deviation of the yields obtained using different PDF sets was used as the absolute uncertainty due to

PDF. The relative uncertainty is then computed by dividing the standard deviation by the mean yield.

- Comparison of the nominal AMC@NLO MC samples to alternative SHERPA (v2.2) samples produced at leading-order with one extra parton in the matrix element for ttW and 2 extra partons for ttZ [58]. The yield comparison for all SRs is shown in Table 7.14, with negligible differences in some SRs and up to 28% in the worst case.

As a result of these studies, the total theory uncertainty for these processes is at the level of 15-35% in the signal and validation regions used in the analysis.

Diboson $WZ, ZZ, W^\pm W^\pm$ production

The theoretical uncertainties on the WZ and ZZ processes are evaluated by several variations added in quadrature :

- Normalization and factorization scales varied independently up and down by a factor of two from the central scale choice. The largest deviation with respect to the nominal is used as the symmetric uncertainty.
- The standard deviation of the yields obtained using different PDF sets was used as the absolute uncertainty due to PDF. The relative uncertainty is then computed by dividing the standard deviation by the mean yield.
- Re-summation scale varied up and down by a factor of two from the nominal value.
- The scale for calculating the overlap between jets from the matrix element and the parton shower is varied from the nominal value of 20 GeV down to

15 GeV and up to 30 GeV. The largest deviation with respect to the nominal is used as the symmetric uncertainty due to matrix element matching.

- An alternative recoil scheme is considered to estimate the uncertainty associated with mis-modeling of jet multiplicities larger than three.

Based on these studies and the cross-section uncertainties, the total theory uncertainty for these processes is at the level of 25-40% in the signal and validation regions used in the analysis.

No theoretical uncertainties have been evaluated specifically for the $W^\pm W^\pm jj$ process, to which we assign the same uncertainties as for WZ , by lack of a better choice. But it should be noted that contributions from this process are minor in the SRs and typically smaller than those from WZ and ZZ .

Other rare processes

A conservative 50% uncertainty is assigned on the summed contributions of all these processes ($t\bar{t}H$, tZ , tWZ , $t\bar{t}t\bar{t}$, $t\bar{t}WW$, $t\bar{t}WZ$, WH , ZH , VVV), which is generally quite larger than the uncertainties on their inclusive production cross-sections, and assumes a similar level of mis-modelling as for diboson or $t\bar{t}V$ processes.

7.4.2 Experimental Uncertainties

Uncertainties associated with the measurement and reconstruction of the physics objects used in the analysis (leptons, jets, etc.) must be accounted for when interpreting the results. The systematic uncertainties from the data-driven method

have already been discussed in Section 7.3. In fact, these data-driven backgrounds are affected by the same systematic uncertainty as in data to which they are being compared to. As a result, only systematic uncertainties on backgrounds estimated with MC simulation and detector simulation needs to be considered. The uncertainties considered for the analysis and recommended by the ATLAS SUSY group are :

Jet energy scale (JES)

In order to account for inefficiencies in the calorimeter cells and the varying response to charged and neutral particles passing through them, the energies of the jets used in this analysis were corrected. The calibration procedure uses a combination of simulation and test beam and in situ data [68] with an uncertainty correlated between all events. As a result, all distributions used in the final result are produced with the nominal calibration as well as an up and down variation of the jet energy scale (in a fully correlated way) by the $\pm 1\sigma$ uncertainty of each nuisance parameter. A combined version of several independent sources contributing to the calibration was used in the analysis to reduce the number of nuisance variables in the fitting procedure.

Jet energy resolution (JER)

An extra p_T smearing is added to the jets based on their p_T and η to account for a possible underestimate of the jet energy resolution in the MC simulation. A systematic uncertainty is considered to account for this defect on the final result. The JER in data has previously been estimated by ATLAS in dijet events.

Jet vertex tagger The uncertainties account for the residual contamination from pile-up jets after pile-up suppression and the MC generator choice [69].

Flavor tagging The MC simulation does not reproduce correctly the b -tagging, charm identification, and light jet reject efficiencies of the detector. A $t\bar{t}$ MC simulation and di-jet measurements are used to derive correction factors to be applied to MC simulation [71, 72]. These correction factors are then varied within their uncertainties to produce up and down variations.

Lepton energy scale, resolution, and Identification efficiencies Similar to the case of jets, electrons and muons have also corresponding energy scale and resolution systematic uncertainties. Corrections are also applied to take into account any variations in the identification efficiency in the detector and its simulation [26, 73, 26].

E_T^{miss} soft term uncertainties The main effect come from the hard object uncertainties (most notably JES and JER) that are propagated to the E_T^{miss} .

Pileup re-weighting This uncertainty is obtained by re-scaling the μ value in data by 1.00 and 1/1.18, covering the full difference between applying and not-applying the nominal μ correction of 1/1.09, as well as effects resulting from uncertainties on the luminosity measurements, which are expected to dominate.

Luminosity The integrated luminosities in data corresponds to 3.2 fb^{-1} and 32.9 fb^{-1} for 2015 and 2016 respectively. The combined luminosity error for 2015 and 2016 is 3.2%, assuming partially correlated uncertainties in 2015 and 2016.

Trigger To account for any differences between the trigger efficiency in simulation and data, corrections factors are derived to correct for them. Uncertainties on the correction factors as well as inefficiencies related to the plateau of the trigger are propagated to the final result.

The uncertainty on the beam energy is neglected. All the experimental uncertainties are applied also on the signal samples when computing exclusion limits on SUSY scenarios.

All of these uncertainties are fed into the fitting and limit setting machinery by treating them as uncorrelated uncertainties, and thus treated independently.

Validation Region	$N_{\text{leptons}}^{\text{signal}}$	$N_{b\text{-jets}}$	N_{jets}	p_T^{jet} [GeV]	E_T^{miss} [GeV]	m_{eff} [GeV]	Other
$t\bar{t}W$	= 2SS	≥ 1	$\geq 4 (e^\pm e^\pm, e^\pm \mu^\pm)$ $\geq 3 (\mu^\pm \mu^\pm)$	> 40 > 25	> 45	> 550	$p_T^{\ell_2} > 40 \text{ GeV}$ $\sum p_T^{b\text{-jet}} / \sum p_T^{\text{jet}} > 0.25$
$t\bar{t}Z$	≥ 3 $\geq 1 \text{ SFOS pair}$	≥ 1	≥ 3	> 35	–	> 450	$81 < m_{\text{SFOS}} < 101 \text{ GeV}$
$WZ4j$	= 3	= 0	≥ 4	> 25	–	> 450	$E_T^{\text{miss}} / \sum p_T^\ell < 0.7$
$WZ5j$	= 3	= 0	≥ 5	> 25	–	> 450	$E_T^{\text{miss}} / \sum p_T^\ell < 0.7$
$W^\pm W^\pm jj$	= 2SS	= 0	≥ 2	> 50	> 55	> 650	veto $81 < m_{e^\pm e^\pm} < 101 \text{ GeV}$ $p_T^{\ell_2} > 30 \text{ GeV}$ $\Delta R_\eta(\ell_{1,2}, j) > 0.7$ $\Delta R_\eta(\ell_1, \ell_2) > 1.3$
All VRs	Veto events belonging to any SR						

TABLE 7.3: Summary of the event selection in the validation regions (VRs).

Requirements are placed on the number of signal leptons ($N_{\text{leptons}}^{\text{signal}}$), the number of b -jets with $p_T > 20 \text{ GeV}$ ($N_{b\text{-jets}}$) or the number of jets (N_{jets}) above a certain p_T threshold (p_T^{jet}). The two leading- p_T leptons are referred to as $\ell_{1,2}$ with decreasing p_T . Additional requirements are set on E_T^{miss} , m_{eff} , the invariant mass of the two leading electrons $m_{e^\pm e^\pm}$, the presence of SS leptons or a pair of same-flavour opposite-sign leptons (SFOS) and its invariant mass m_{SFOS} . A minimum angular separation between the leptons and the jets ($\Delta R_\eta(\ell_{1,2}, j)$) and between the two leptons ($\Delta R_\eta(\ell_1, \ell_2)$) is imposed in the $W^\pm W^\pm jj$ VR. For the two WZ VRs the selection also relies on the ratio of the E_T^{miss} in the event to the sum of p_T of all signal leptons p_T ($E_T^{\text{miss}} / \sum p_T^\ell$). The ratio of the scalar sum of the p_T of all b -jets to that of all jets in the event ($\sum p_T^{b\text{-jet}} / \sum p_T^{\text{jet}}$) is used in the $t\bar{t}W$ VR selection.

Validation Region	$t\bar{t}W$	$t\bar{t}Z$	$WZ4j$	$WZ5j$	$W^\pm W^\pm jj$
$t\bar{t}Z/\gamma^*$	6.2 ± 0.9	123 ± 17	17.8 ± 3.5	10.1 ± 2.3	1.06 ± 0.22
$t\bar{t}W$	19.0 ± 2.9	1.71 ± 0.27	1.30 ± 0.32	0.45 ± 0.14	4.1 ± 0.8
$t\bar{t}H$	5.8 ± 1.2	3.6 ± 1.8	1.8 ± 0.6	0.96 ± 0.34	0.69 ± 0.14
$4t$	1.02 ± 0.22	0.27 ± 0.14	0.04 ± 0.02	0.03 ± 0.02	0.03 ± 0.02
$W^\pm W^\pm$	0.5 ± 0.4	—	—	—	26 ± 14
WZ	1.4 ± 0.8	29 ± 17	200 ± 110	70 ± 40	27 ± 14
ZZ	0.04 ± 0.03	5.5 ± 3.1	22 ± 12	9 ± 5	0.53 ± 0.30
Rare	2.2 ± 0.5	26 ± 13	7.3 ± 2.1	3.0 ± 1.0	1.8 ± 0.5
Fake/non-prompt leptons	18 ± 16	22 ± 14	49 ± 31	17 ± 12	13 ± 10
Charge-flip	3.4 ± 0.5	—	—	—	1.74 ± 0.22
Total SM background	57 ± 16	212 ± 35	300 ± 130	110 ± 50	77 ± 31
Observed	71	209	257	106	99

TABLE 7.4: The numbers of observed data and expected background events in the validation regions. The rare category is defined in the text. Background categories with yields shown as “—” do not contribute to a given region (e.g. charge flips in three-lepton regions) or their estimates are below 0.01 events. The displayed yields include all statistical and systematic uncertainties.

TABLE 7.5: Muon fake rate measured in data and the associated statistical uncertainty. The systematic uncertainty originating from the subtraction of “backgrounds” with only prompt leptons is also displayed.

$10 < p_T < 12 \text{ GeV}$		$12 < p_T < 14$	
$ \eta < 2.3$	$ \eta > 2.3$	$ \eta < 2.3$	$ \eta > 2.3$
$0.14 \pm 0.01 \pm 0.00$	$0.22 \pm 0.05 \pm 0.00$	$0.11 \pm 0.01 \pm 0.00$	$0.24 \pm 0.06 \pm 0.00$
$14 < p_T < 17$		$17 < p_T < 20 \text{ GeV}$	
$ \eta < 2.3$	$ \eta > 2.3$	$ \eta < 2.3$	$ \eta > 2.3$
$0.12 \pm 0.01 \pm 0.00$	$0.09 \pm 0.05 \pm 0.00$	$0.09 \pm 0.01 \pm 0.00$	$0.21 \pm 0.07 \pm 0.00$
$20 < p_T < 30$	$30 < p_T < 40$	$40 < p_T < 60$	$p_T > 60$
$0.07 \pm 0.02 \pm 0.00$	$0.12 \pm 0.05 \pm 0.01$	$0.16 \pm 0.09 \pm 0.04$	$0.49 \pm 0.10 \pm 0.07$

TABLE 7.6: Additional systematic uncertainty on the muon fake rates, to address variations of the latter in different environments. The table also shows the correction factors and uncertainties applied to final states with $\geq 2 b$ -tagged jets.

p_T	< 14	$14 - 20$	$20 - 30$	$30 - 40$	$40 - 60$	> 60
$\Delta\zeta^{(\text{syst})}$	30%	30%	30%	50%	50% for $H_T < 600$ 70% for $600 < H_T < 1200$ 85% for $H_T > 1200$	
$\frac{\zeta_{\geq 2b}}{\zeta}$	1.2 ± 0.2	1.5 ± 0.5	1.7 ± 0.7	2.0 ± 1.0	1.5 ± 0.5	—

TABLE 7.7: Electron fake rate measured in data and the associated statistical uncertainty. The systematic uncertainty originating from the subtraction of “backgrounds” with only prompt leptons is also displayed.

$10 < p_T < 12$	$12 < p_T < 14$	$14 < p_T < 17$	$17 < p_T < 20$
$0.10 \pm 0.01 \pm 0.00$	$0.10 \pm 0.01 \pm 0.01$	$0.12 \pm 0.01 \pm 0.01$	$0.08 \pm 0.02 \pm 0.00$
$20 < p_T < 25$	$25 < p_T < 30$	$30 < p_T < 40$	$40 > p_T$
$0.07 \pm 0.02 \pm 0.01$	$0.11 \pm 0.03 \pm 0.01$	$0.20 \pm 0.07 \pm 0.03$	$0.25 \pm 0.10 \pm 0.05$

	Electrons			Muons			
	$0 < \eta < 0.8$	$0.8 < \eta < 1.37$	$1.52 < \eta < 2.01$	$0 < \eta < 0.6$	$0.6 < \eta < 1.2$	$1.2 < \eta < 1.8$	$1.8 < \eta < 2.5$
$10 \text{ GeV} < p_T < 15 \text{ GeV}$	0.047	0.063	0.089	0.014	0.010	0.008	0.011
$15 \text{ GeV} < p_T < 20 \text{ GeV}$	0.027	0.042	0.062	0.005	0.006	0.008	0.011
$20 \text{ GeV} < p_T < 25 \text{ GeV}$	0.018	0.031	0.041	0.003	0.006	0.010	0.010
$25 \text{ GeV} < p_T < 30 \text{ GeV}$	0.029	0.024	0.027	0.011	0.015	0.022	0.019
$30 \text{ GeV} < p_T < 35 \text{ GeV}$	0.023	0.021	0.023	0.007	0.009	0.014	0.011
$35 \text{ GeV} < p_T < 40 \text{ GeV}$	0.014	0.018	0.018	0.004	0.004	0.006	0.006
$40 \text{ GeV} < p_T < 50 \text{ GeV}$	0.007	0.010	0.010	0.002	0.001	0.002	0.001
$50 \text{ GeV} < p_T < 60 \text{ GeV}$	0.008	0.010	0.010	0.001	0.001	0.001	0.001
$60 \text{ GeV} < p_T < 70 \text{ GeV}$	0.007	0.010	0.010	0.001	0.001	0.001	0.002
$70 \text{ GeV} < p_T < 80 \text{ GeV}$	0.008	0.011	0.012	0.002	0.001	0.001	0.002
$80 \text{ GeV} < p_T < 120 \text{ GeV}$	0.010	0.010	0.011	0.004	0.002	0.002	0.002
$120 \text{ GeV} < p_T < 150 \text{ GeV}$	0.005	0.005	0.011	0.006	0.005	0.005	0.005
$150 \text{ GeV} < p_T < 200 \text{ GeV}$	0.005	0.002	0.019	0.005	0.005	0.005	0.006

TABLE 7.8: Systematic uncertainties on the measured real lepton efficiency, separating sources affecting the measurement itself (background subtraction, trigger bias, and different methods).

electrons (busy environments)								
$\Delta R(e, jet)$	[0, 0.1]	[0.1, 0.15]	[0.15, 0.2]	[0.2, 0.3]	[0.3, 0.35]	[0.35, 0.4]	[0.4, 0.6]	[0.6, 4]
10 GeV < p_T < 20 GeV	-	-	-	-	-	-	25.31%	6.5%
20 GeV < p_T < 30 GeV	-	-	-	-	-	-	73.37%	10.21%
30 GeV < p_T < 40 GeV	-	-	-	97.71%	48.22%	15.54%	7.29%	0.58%
40 GeV < p_T < 50 GeV	-	-	-	52.81%	22.80%	16.73%	7.68%	1.10%
50 GeV < p_T < 60 GeV	-	-	-	29.96%	21.49%	20.23%	6.99%	2.78%
60 GeV < p_T < 80 GeV	-	-	55.89%	24.31%	17.40%	24.77%	6.20%	2.87%
80 GeV < p_T < 150 GeV	-	57.52%	30.24%	16.45%	12.73%	20.92%	4.44%	2.73%
150 GeV < p_T < 200 GeV	88.54%	40.16%	19.34%	8.45%	14.66%	16.57%	2.57%	1.90%

muons (busy environments)								
$\Delta R(\mu, jet)$	[0, 0.1]	[0.1, 0.15]	[0.15, 0.2]	[0.2, 0.3]	[0.3, 0.35]	[0.35, 0.4]	[0.4, 0.6]	[0.6, 4]
10 GeV < p_T < 20 GeV	-	-	-	-	-	-	33.59%	5.18%
20 GeV < p_T < 30 GeV	-	-	-	-	-	-	82.34%	22.27%
30 GeV < p_T < 40 GeV	-	-	-	98.54%	56.36%	31.89%	14.22%	2.24%
40 GeV < p_T < 50 GeV	-	-	-	53.10%	21.33%	13.90%	6.81%	1.45%
50 GeV < p_T < 60 GeV	-	-	-	24.98%	13.72%	9.62%	3.83%	0.79%
60 GeV < p_T < 80 GeV	-	-	44.41%	13.75%	6.14%	4.76%	2.04%	0.15%
80 GeV < p_T < 150 GeV	-	29.94%	7.14%	3.16%	1.30%	1.04%	0.07%	0.57%
150 GeV < p_T < 200 GeV	82.26%	4.14%	1.02%	0.17%	0.29%	0.62%	1.02%	1.13%

TABLE 7.9: Systematic uncertainties on the measured real lepton efficiency, due to the extrapolation to busy environments using $\tilde{g} \rightarrow t\bar{t}\tilde{\chi}_1^0$ events.

TABLE 7.10: Expected yields for background processes with fake leptons, in the signal regions with a global correction factor that represents the ratio of weighted $t\bar{t}$ to raw MC $t\bar{t}$ with a global uncertainty that includes : fit uncertainty, theory uncertainties on $t\bar{t}$, comparison of different showers. The fraction of the systematic uncertainty from the comparison between two showers (Pythia and Sherpa) is also shown.

Region	MC Template method	Global correction	Shower systematic
Rpc2L0bH	$1.00 \pm 0.96 \pm 0.81$	2.80 ± 2.10	74%
Rpc2L0bS	$1.68 \pm 1.02 \pm 1.26$	2.89 ± 1.97	65%
Rpc2L1bH	$2.07 \pm 0.63 \pm 1.56$	1.22 ± 1.14	34%
Rpc2L1bS	$2.33 \pm 1.17 \pm 2.10$	1.83 ± 1.42	81%
Rpc2L2bH	< 0.5	0 ± 0	0%
Rpc2L2bS	$0.41 \pm 0.33 \pm 0.45$	1.47 ± 1.12	73%
Rpc2Lsoft1b	$2.48 \pm 1.32 \pm 1.86$	1.59 ± 1.31	68%
Rpc2Lsoft2b	$1.66 \pm 0.66 \pm 1.28$	1.72 ± 1.29	54%
Rpc3L0bH	< 0.5	0 ± 0	0%
Rpc3L0bS	$0.21 \pm 0.15 \pm 0.16$	2.90 ± 2.20	71%
Rpc3L1bH	$0.42 \pm 0.29 \pm 0.32$	1.59 ± 1.25	59%
Rpc3L1bS	$3.55 \pm 1.80 \pm 2.76$	1.76 ± 1.32	67%
Rpc3LSS1b	$0.90 \pm 0.14 \pm 0.69$	2.34 ± 1.44	56%

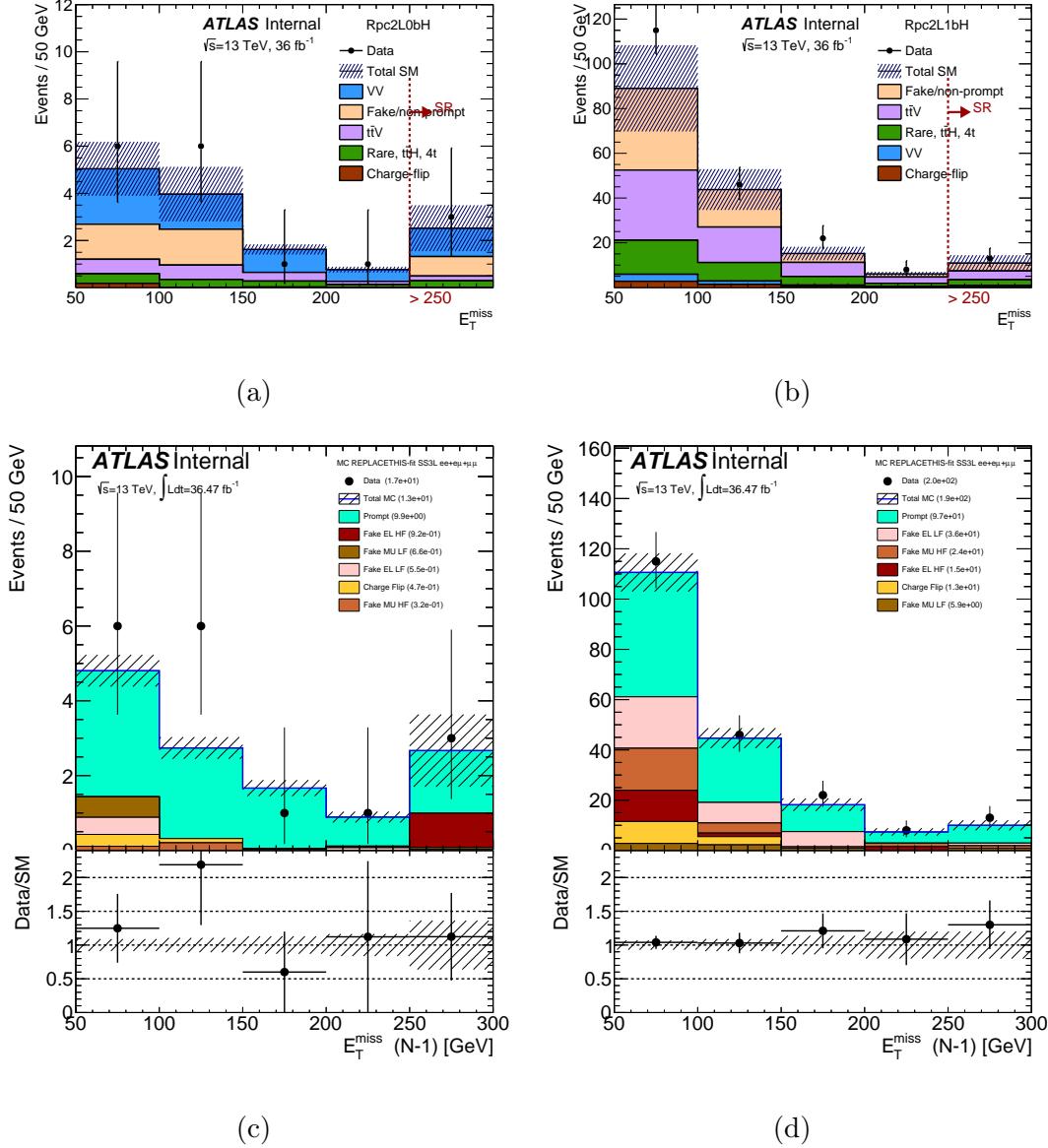


FIGURE 7.23: Missing transverse momentum distributions after (7.23a–7.23c) Rpc2L0bH and (7.23b–7.23d) Rpc2L1bH selection, except the E_T^{miss} requirement. Estimates with the matrix method are in the upper plots (7.23a–7.23a) while estimates with the MC template method are in the lower plots (7.23c–7.23c). The results in the signal regions are shown in the last (inclusive) bin of each plot. The statistical uncertainties in the background prediction are included in the uncertainty band, as well as the full systematic uncertainties for backgrounds with fake or non-prompt leptons, or charge-flip.

TABLE 7.11: Expected yields for background processes with fake leptons, in the signal regions shown for 36 fb^{-1} . Estimates from the matrix method and the MC template method are shown along with the retained estimates. Uncertainties include all statistical and systematic sources for the nominal estimate.

Region	Matrix method	Template method	Retained estimate
Rpc2L0bH	$0.83 \pm 0.56 \pm 0.74$	$1.00 \pm 0.96 \pm 0.81$	$0.87 \pm 0.48 \pm 0.76$
Rpc2L0bS	$1.51 \pm 0.60 \pm 0.66$	$1.68 \pm 1.02 \pm 1.26$	$1.55 \pm 0.52 \pm 0.81$
Rpc2L1bH	$3.54 \pm 1.62 \pm 3.12$	$2.07 \pm 0.63 \pm 1.56$	$2.26 \pm 0.59 \pm 1.76$
Rpc2L1bS	$2.65 \pm 1.21 \pm 1.89$	$02.33 \pm 01.17 \pm 02.10$	$2.48 \pm 0.84 \pm 2.00$
Rpc2L2bH	$-0.11 \pm 0.11 \pm 0.18$	< 0.5	$0.15 \pm 0.15 \pm 0.00$
Rpc2L2bS	$1.31 \pm 1.07 \pm 1.65$	$0.41 \pm 0.33 \pm 0.45$	$0.49 \pm 0.32 \pm 0.55$
Rpc2Lsoft1b	$4.75 \pm 1.42 \pm 2.64$	$2.48 \pm 1.32 \pm 1.86$	$3.53 \pm 0.97 \pm 2.22$
Rpc2Lsoft2b	$1.91 \pm 1.18 \pm 1.63$	$1.66 \pm 0.66 \pm 1.28$	$1.72 \pm 0.58 \pm 1.36$
Rpc3L0bH	$-0.01 \pm 0.11 \pm 0.10$	< 0.5	$0.15 \pm 0.15 \pm 0.00$
Rpc3L0bS	$2.31 \pm 1.50 \pm 2.63$	$0.21 \pm 0.15 \pm 0.16$	$0.23 \pm 0.15 \pm 0.18$
Rpc3L1bH	$0.57 \pm 0.43 \pm 0.50$	$0.42 \pm 0.29 \pm 0.32$	$0.47 \pm 0.24 \pm 0.38$
Rpc3L1bS	$4.94 \pm 1.83 \pm 2.96$	$3.55 \pm 1.80 \pm 2.76$	$4.23 \pm 1.28 \pm 2.86$
Rpc3LSS1b	$-0.18 \pm 1.24 \pm 2.85$	$0.90 \pm 0.14 \pm 0.69$	$0.89 \pm 0.14 \pm 0.72$

TABLE 7.12: Expected yields for background processes with charge-flipped electrons, in the signal regions shown for 36 fb^{-1} . Estimates from the likelihood method and the MC template method are shown. Uncertainties include all statistical and systematic sources. Charge-flip processes do not contribute to signal regions which require ≥ 3 leptons.

Region	Weighted OS data	Template method
Rpc2L0bH	$0.01 \pm 0.00 \pm 0.00$	< 0.4
Rpc2L0bS	$0.05 \pm 0.01 \pm 0.01$	$00.02 \pm 00.02 \pm 00.00$
Rpc2L1bH	$0.25 \pm 0.03 \pm 0.04$	$00.21 \pm 00.32 \pm 00.16$
Rpc2L1bS	$0.25 \pm 0.02 \pm 0.04$	$00.35 \pm 00.37 \pm 00.26$
Rpc2L2bH	$0.02 \pm 0.01 \pm 0.00$	< 0.4
Rpc2L2bS	$0.10 \pm 0.01 \pm 0.02$	< 0.4
Rpc2Lsoft1b	$0.08 \pm 0.01 \pm 0.02$	< 0.4
Rpc2Lsoft2b	$0.08 \pm 0.01 \pm 0.02$	< 0.4
Rpc3LSS1b	$0.39 \pm 0.03 \pm 0.07$	$00.81 \pm 00.53 \pm 00.34$

TABLE 7.13: Comparison of expected yields for background processes with fake leptons, in the validation regions, shown for 36 fb^{-1} between the data driven (DD) estimates and the MC template method (MC) estimates.

	VR- $t\bar{t}W$	VR- $t\bar{t}Z$	VR- $WZ4j$	VR- $WZ5j$	VR- $W^\pm W^\pm$
Fakes DD	$23 \pm 5 \pm 24$	$30 \pm 4 \pm 14$	$53 \pm 6 \pm 27$	$21 \pm 4 \pm 10$	$14 \pm 3 \pm 10$
Fakes MC	$14 \pm 4 \pm 10$	$18 \pm 3 \pm 13$	$46 \pm 5 \pm 34$	$16 \pm 2 \pm 12$	$13 \pm 2 \pm 10$
Combined	$18 \pm 3 \pm 15$	$22 \pm 2 \pm 13$	$49 \pm 4 \pm 30$	$17 \pm 2 \pm 12$	$13 \pm 2 \pm 10$
Charge-flip DD	$3.4 \pm 0.1 \pm 0.5$	–	–	–	$1.7 \pm 0.1 \pm 0.2$
Charge-flip MC	$3.8 \pm 1.0 \pm 1.9$	–	–	–	$1.0 \pm 0.3 \pm 0.2$

TABLE 7.14: Comparison of the event yields for the $t\bar{t}V$ background processes between AMC@NLO (default generator) and SHERPA in the SRs, as well as their relative difference.

SR	SHERPA	aMCATNLO	Relative diff.
Rpc2L0bH	0.25 ± 0.03	0.20 ± 0.05	25%
Rpc2L0bS	0.60 ± 0.06	0.82 ± 0.10	-26%
Rpc2L1bH	3.84 ± 0.14	3.86 ± 0.20	<1%
Rpc2L1bS	3.55 ± 0.13	3.94 ± 0.20	-9%
Rpc2L2bH	0.35 ± 0.04	0.41 ± 0.05	-14%
Rpc2L2bS	1.57 ± 0.08	1.57 ± 0.12	<1%
Rpc2Lsoft1b	1.01 ± 0.07	1.24 ± 0.11	-18%
Rpc2Lsoft2b	1.13 ± 0.07	1.15 ± 0.10	-1%
Rpc3L0bH	0.23 ± 0.02	0.18 ± 0.04	27%
Rpc3L0bS	0.90 ± 0.05	0.99 ± 0.09	-9%
Rpc3L1bH	1.54 ± 0.08	1.52 ± 0.11	1%
Rpc3L1bS	6.95 ± 0.16	7.02 ± 0.23	<1%
Rpc3LSS1b	0.00 ± 0.00	0.00 ± 0.00	-

Chapitre 8

Statistical Treatment

The goal of the analysis is to maximize the information that can be extracted from comparing the observed data to the background prediction in the signal regions designed to search for new physics topologies. Statistical tools are essential to tell in the most powerful way and to the best of our knowledge if there is a new physics signal beyond what is already known in the observed data. At the same time, it is important to properly treat the systematic uncertainties associated with the complexity of the experimental apparatus (ATLAS detector) and the background predictions when presenting an interpretation of the results. This chapter describes the statistical methodology employed to test the compatibility between data and prediction while taking into account the systematic uncertainties. The analysis' possible outcomes are represented by a likelihood function that combines observations, predictions, and associated uncertainties. At which point the hypothesis testing is performed with the corresponding one-sided profile likelihood ratio [81], and upper limits are provided as one-sided 95% confidence level intervals in the CL_s formalism [82]. The statistical tool used to perform the quantification of the significance of hypothetical excesses seen in data or upper limits setting on new physics contributions as implemented in this analysis will be described in this chapter.

8.1 Likelihood Function

The likelihood for a set of parameters $(\mu, \boldsymbol{\theta})$ given all the data that might have been observed \mathbf{X} is the probability of observing the data given the parameters

$$\mathcal{L}(\mu, \boldsymbol{\theta} | \mathbf{X}) = \Pr(\mathbf{X} | \mu, \boldsymbol{\theta}). \quad (8.1)$$

The data \mathbf{X} includes observation in the signal regions as well as other auxiliary experiments such as control regions used to constrain backgrounds. In this analysis, hypothesis tests are performed on one signal region at the time, single-binned, and without control regions. As a result, the observed data \mathbf{X} has one-dimensional component with value X representing the count of events in the signal region. The first parameter of interest represents the ‘strength’ of the signal process $\mu > 0$ that will increase the number of expected events in the signal region given that the signal of the new physics model tested is present. In practice, the signal strength μ is used to scale the nominal expected cross section for the signal process, or the number of expected signal events s . Thus, the predicted background will be of the form $\mu s + \sum_i b_i$ where b_i represents the standard model background processes expected to contribute to the signal region. The parameter $\boldsymbol{\theta}$ refers to the nuisance parameters used to parametrize the systematic uncertainties (luminosity, JES, JER, etc.)¹. Thus, the likelihood is built as the product of a Poisson probability density function describing the observed number of events in the signal region and

1. The parameters represented by $\boldsymbol{\theta}$ are called nuisance parameters since the aim is not to set a limit on them.

Gaussian distributions for each of the sources of systematic uncertainties. The likelihood takes the simple form

$$\Pr(X, \boldsymbol{\theta}^0 | \mu, \boldsymbol{\theta}) = \mathcal{P}\left(X \mid \mu s(\boldsymbol{\theta}) + \sum_i b_i(\boldsymbol{\theta})\right) \times \prod_j \mathcal{G}(\theta^0 | \theta_j) \quad (8.2)$$

where $\mathcal{P}(X | \nu) = e^{-\nu} \nu^X / X!$ and $\mathcal{G}(\theta) = \frac{1}{\sqrt{2\pi\sigma_\theta^2}} e^{-\frac{(\theta-\theta^0)^2}{2\sigma_\theta^2}}$ are the Poisson and Gaussian probability density functions. Both signal and backgrounds depend on the nuisance parameter $\boldsymbol{\theta}$ which controls all independent sources of uncertainty and will be *profiled* (or constrained) in the CL_s procedure described next. Correlations of a given nuisance parameter between the different sources of backgrounds and the signal are taken into account when relevant. To give an example, the luminosity uncertainty will have a mean as the luminosity central value $\theta_{\text{lumi}}^{(0)}$ and the width as an experimentally determined uncertainty $\sigma_{\theta_{\text{lumi}}}$. When evaluating the effect of the luminosity uncertainty on the likelihood function, all terms involving the nuisance parameter θ_{lumi} will be scaled in the same way since luminosity is correlated across all backgrounds and signal.

The likelihood function described in this section is used in a fit to data by the maximum likelihood method that aims at finding the value of the signal strength μ that makes the likelihood a maximum. The procedure relies on an iterative minimization algorithm implemented in MINUIT [83] and accessed by *RooFit* [84] within ROOT, a high energy physics data analysis framework [85]. The final uncertainties on the nuisance parameters in $\boldsymbol{\theta}$, constrained by the fit, are obtained from the covariance matrix of these parameters.

8.2 Limit Setting Procedure

8.2.1 Profile Likelihood Ratio

The procedure for setting exclusion limits using the likelihood function (Eq.8.2) relies on a profile-likelihood-ratio test [81]. The null hypothesis considered is that of background only with $\mu = 0$ and the alternate hypothesis is the presence of a signal with strength $\mu > 0$. According to the Neyman-Pearson Lemma [86], the most powerful test when performing a hypothesis test between two simple hypotheses, $\mu = 0$ and $\mu > 0$, is the profile-likelihood-ratio test, which rejects $\mu = 0$ in favor of $\mu > 0$. The profile-likelihood-ratio test q_μ is defined to be

$$q_\mu(\mathbf{X}) = \begin{cases} -2 \ln \left(\frac{\mathcal{L}(\mu, \hat{\theta}(\mu) | \mathbf{X})}{\mathcal{L}(\hat{\mu}, \hat{\theta} | \mathbf{X})} \right), & \text{if } \mu > \hat{\mu} \\ 0, & \text{otherwise} \end{cases} \quad (8.3)$$

For a given observation \mathbf{X} , the parameters $\hat{\mu}$ and $\boldsymbol{\theta}$ are obtained from maximizing the likelihood. If the value of μ is also fixed, the set of values $\hat{\boldsymbol{\theta}}$ corresponds to the set of values that maximize the likelihood for that particular value of μ . The variable q_μ is always positive ($q_\mu \geq 0$). Qualitatively, small values of q_μ correspond to a better compatibility between the observed data and the tested hypothesis with a signal strength μ , while large values indicate a very improbable signal hypothesis. The condition on the sign of $\mu - \hat{\mu}$ is necessary in order not to interpret an upward fluctuation of data ($X > \mu s + b$) as incompatible with the tested hypothesis. The form of q_μ is motivated by Wald and Wilk's [87, 88] approximation formulas in the large sample limit where the distributions of q_μ also

become independent of nuisance parameters as it will be discussed next. The test statistic for discovery will try to reject the background only hypothesis ($\mu = 0$). Thus Eq.8.3 becomes

$$q_0(\mathbf{X}) = \begin{cases} -2 \ln \left(\frac{\mathcal{L}(0, \hat{\theta}(0) | \mathbf{X})}{\mathcal{L}(\hat{\mu}, \hat{\theta} | \mathbf{X})} \right), & \text{if } \hat{\mu} < 0 \\ 0, & \text{otherwise} \end{cases} \quad (8.4)$$

8.2.2 The p -value

At this stage, a test statistic q_μ has been constructed to distinguish between the hypothesis that the data contains signal and background $\mu > 0$ and that of background only $\mu = 0$. To illustrate the limit setting procedure, we consider distributions of the test statistic under each hypothesis : $f(q_\mu | \mu)$ for $\mu > 0$ and $f(q_0 | 0)$ for $\mu = 0$. These distributions are shown in Figure 8.1a and details on how to obtain their functional forms will be discussed later.

Given that the actual observed data leads to a test variable $q_\mu^{(\text{obs})}$, it is possible to quantify the level of discrepancy between the observed data and the tested hypothesis ($\mu > 0$ or $\mu = 0$) using a p -value. The p -value of the signal hypothesis ($\mu > 0$) is then defined as the probability, under assumption of the signal hypothesis, to find a value of q_μ with equal or lesser compatibility with the signal model considered relative to what is found with $q_\mu^{(\text{obs})}$. In other words, higher values of q_μ indicate an increasing disagreement between data and the signal model. The mathematical expression of the p -value with $\mu > 0$ is taken as the probability to

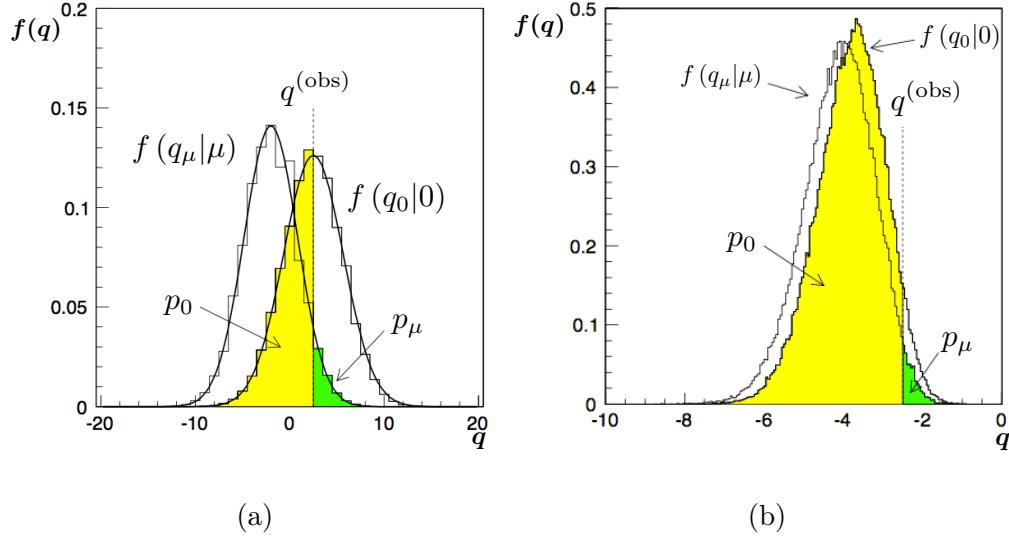


FIGURE 8.1: Distributions of the test variable q_μ under the $\mu > 0$ and $\mu = 0$ hypotheses : (a) typical case, (b) case where there is very little sensitivity to the signal model

find q_μ greater than or equal $q_\mu^{(\text{obs})}$, under the signal hypothesis² is given by

$$p_\mu = \Pr(q_\mu \geq q_\mu^{(\text{obs})} | \mu) = \int_{q_\mu^{(\text{obs})}}^{\infty} f(q_\mu | \mu) dq_\mu \quad (8.5)$$

where $q_\mu^{(\text{obs})}$ is the value of the statistic test observed in data and the function f denotes the probability distribution function of q_μ under the signal hypothesis. Similarly, the p -value of the background-only hypothesis with $\mu = 0$ takes the form of

$$p_0 = \Pr(q_0 \geq q_0^{(\text{obs})} | 0) = \int_{-\infty}^{q_0^{(\text{obs})}} f(q_0 | 0) dq_0 \quad (8.6)$$

and can be interpreted as the probability of the observation to be consistent with the background only hypothesis : the smaller the p_0 value is, the less compatible data is with the background only hypothesis. In order to claim a discovery in

2. Note that the background-only distribution $f(q_0 | 0)$ in the example given in Figure 8.1a is shifted to the right.

particle physics, the background-only hypothesis must be rejected using the p_0 -value. However, there is an ambiguity related to how small the p_0 -value needs to be, before declaring a discovery. The problem can also be formulated in terms of significance. Assuming a Gaussian distributed variable, how many Z standard deviations σ above the mean are required to cover an upper-tail probability equal to p expressed as

$$Z = \Phi^{-1}(1 - p) \quad (8.7)$$

where Φ is the Gaussian cumulative distribution. The particle physics community considers a discovery if $Z = 5$, commonly used “five sigma excess” statement, which corresponds to $p_0 = 2.87 \times 10^{-7}$ or one in 3.5 million probability for the background to be as extreme as the observation. In order to announce evidence for a new particle, a significance of $Z = 3$ or $p_0 = 1.35 \times 10^{-3}$ is required.

8.2.3 The CL_s Prescription

The next step is to define a confidence interval (CI) that includes the parameter μ at a specified confidence level (CL). In other words, instead of estimating the parameter μ by a single value, an interval CI likely to include the parameter is given. The CL provides a quantitative statement on how likely the interval CI is to contain the parameter μ . If given the measurement of a parameter μ_{meas} , we deduce that there is a 95% CI $[\mu_1, \mu_2]$ means that in an ensemble of experiments 95% of the obtained CIs will contain the true value of μ . The upper limit is simply the case where the 95% CI is $[0, \mu_{up}]$: In an ensemble of experiments 95% of the obtained CIs will contain the true value of μ , including $\mu = 0$. The conclusion is

that $\mu < \mu_{up}$ at the 95% CL or that μ_{up} is an upper limit.

Applying this concept to the problem at hand where we want to place an upper limit on the expected number of signal events S ($S = \mu s$, s being the expected number of signal events for $\mu = 1$) in one or more signal regions. The CL is obtained from a standard statistical test of the signal model ($\mu > 0$) which can establish the exclusion of the signal model at confidence level $1 - \alpha = 95\%$ if

$$CL_{s+b} = p_\mu < \alpha \quad (8.8)$$

where $\alpha = 0.05$. Since the result section will present CL values in terms of the expected number of signal events from beyond the standard model processes, we continue this discussion of estimating an upper limit on S rather than μ . Thus, a CI at confidence level $CL = 1 - \alpha$ for the expected number of signal events S can be constructed from those values of S (or μ) that are not excluded, and the upper limit $S_{up}^{1-\alpha}$ is the largest value of S not excluded. By construction, the interval $[0, S_{up}^{95}]$ will cover the expected number of signal events S with a probability of at least 95%, regardless of the value of S .

An anomaly arises with the CL_{s+b} prescription when the number of expected signal events is much less than that of the background and the data observation had a downward fluctuation below the expected background. The procedure will lead to excluding, with probability close to α , hypotheses to which the experiment has no sensitivity. For $\alpha = 5\%$, it means that one out of twenty tests for different signal models where one has no sensitivity will result in exclusion. In fact, the desired behavior of the exclusion probability in this case is to approach zero rather

than α . This scenario is illustrated in Figure 8.1b where the distribution of q_μ under both the signal and background-only hypotheses are almost similar. To remedy this problem, a different procedure is used where a model is regarded as excluded if

$$CL_s = \frac{p_\mu}{1 - p_0} < \alpha. \quad (8.9)$$

In this form, the p -value is penalized by dividing by $1 - p_0$. If the distribution of q_μ under signal or background-only hypotheses are widely separated, then the quantity $1 - p_0$ is close to unity which recovers the CL_{s+b} value. However, if the distributions of both hypotheses are similar, due to the lack of sensitivity, $1 - p_0$ becomes smaller and the CL_{s+b} value is increased more leading to a weaker upper limit. Similar to the case of CL_{s+b} , the upper limit on S_{up}^{95} is taken as the largest value of the parameter S not excluded.

8.2.4 Approximate Sampling Distributions

The remaining task is to determine the sampling distributions $f(q_0|0)$ and $f(q_\mu|\mu)$ needed to compute the p -values used in the case of discovery and setting upper limits, respectively. These distributions do not have an analytic form but can be obtained from pseudo-experiments or asymptotic approximations. The pseudo-experiments are more accurate than the asymptotic approximations since a large number of datasets are generated which are drawn from a distribution that is consistent with those observed. However, for a complex likelihood function where the procedure of generating pseudo-datasets needs to be repeated many times for each parameter point of each model being considered, the pseudo-data

set method is computing intensive and not practical. For this reason, the current analysis uses asymptotic formulae to approximate q_μ and shown to be valid in the large sample size limit [81]. The approximation is based on an important result by Wilks [87] and Wald [88] who showed that for a single parameter of interest,

$$q_\mu = \begin{cases} \frac{(\mu - \bar{\mu})^2}{\sigma^2} + \mathcal{O}\left(\frac{1}{\sqrt{N}}\right), & \text{if } \mu > \hat{\mu} \\ 0, & \text{otherwise.} \end{cases} \quad (8.10)$$

where μ is a Gaussian distribution with a mean $\bar{\mu}$ and standard deviation σ , and a sample size N . In the case where $\mu = \hat{\mu}$, the test statistic q_μ follows a χ^2 distribution with one degree of freedom. The variance σ^2 is obtained from an artificial data set called the “Asimov data set”³ that verifies $X_A = \bar{\mu}s + b$. From Eq.8.10, the variance is then $\sigma^2 = (\mu - \bar{\mu})^2 / q_{\mu,A}$, where $q_{\mu,A}$ is evaluated from the exact expression of q_μ using the Asimov data set.

The results obtained from the asymptotic approximations have been compared to exclusion limits obtained with a limited number of pseudo-experiments. A reasonable agreement has been observed which validated the use of the asymptotic formalism to obtain the exclusion limits on the different models.

8.3 Statistical Implementation

The statistical interpretations of the observations are performed with the **HistFitter** framework [90], commonly in use within the SUSY ATLAS working

3. Inspired from the short story *Franchise*[89] by Isaac Asimov that entails replacing a single voter that represents the entire electorate population in an election. Similarly, an ensemble of pseudo-experiments can be replaced by a single representative data set.

group. It consists of a user-friendly abstraction layer to the `HistFactory` and `RooFit` frameworks [91, 92], to which it delegates the tasks of building a probabilistic representation of the analysis, and performing hypothesis tests. `HistFitter` also provides several related utilities, such as the creation of summary tables of yields and uncertainties, or the creation of exclusion plots such as those presented in Chapter 9.

The likelihood function implemented in `HistFitter` is built as the product of a Poisson probability density function describing the observed number of events in the signal region and, to constrain the nuisance parameters associated with the systematic uncertainties, Gaussian distributions whose widths correspond to the sizes of these uncertainties ; Poisson distributions are used instead for MC simulation statistical uncertainties. Correlations of a given nuisance parameter between the backgrounds and the signal are taken into account when relevant. The hypothesis tests are performed for each of the signal regions independently.

Chapitre 9

Results and Interpretation

9.1 Predictions and Observations

Observed data (36.1 fb^{-1}) and predicted SM backgrounds event yields are compared in Fig. 9.2 for signal regions addressing R -parity-conserving signal scenarios, for events satisfying the SR requirements except the E_T^{miss} cut. Adjustments to the selections are made :

- SRs involving an effective mass cut : it is relaxed together with the E_T^{miss} cut to avoid indirectly tightening requirements on the visible leptonic and hadronic contributions to m_{eff} for low values of E_T^{miss} ; the m_{eff} requirement is therefore changed to :

$$m_{\text{eff}} > (m_{\text{eff}}^{\text{SR cut}}) - \max \left[(E_T^{\text{miss, SR cut}}) - E_T^{\text{miss}}, 0 \right]$$

- SRs involving a cut on the $E_T^{\text{miss}}/m_{\text{eff}}$ ratio : it is relaxed together with the E_T^{miss} cut to allow populating the low E_T^{miss} tail of the distributions :

$$\frac{E_T^{\text{miss}}}{m_{\text{eff}}} > \left(\frac{E_T^{\text{miss}}}{m_{\text{eff}}} \right)^{\text{SR cut}} \times \frac{E_T^{\text{miss}}}{\left(E_T^{\text{miss, SR cut}} \right)^{\text{SR cut}}}$$

in addition, an upper cut on m_{eff} is added.

Figure 9.3 shows the event yields for data and the expected background contributions in all signal regions. Detailed information about the yields can be found in Table 9.1. In all 13 SRs the number of observed data events is consistent with the expected background within the uncertainties. The contributions listed

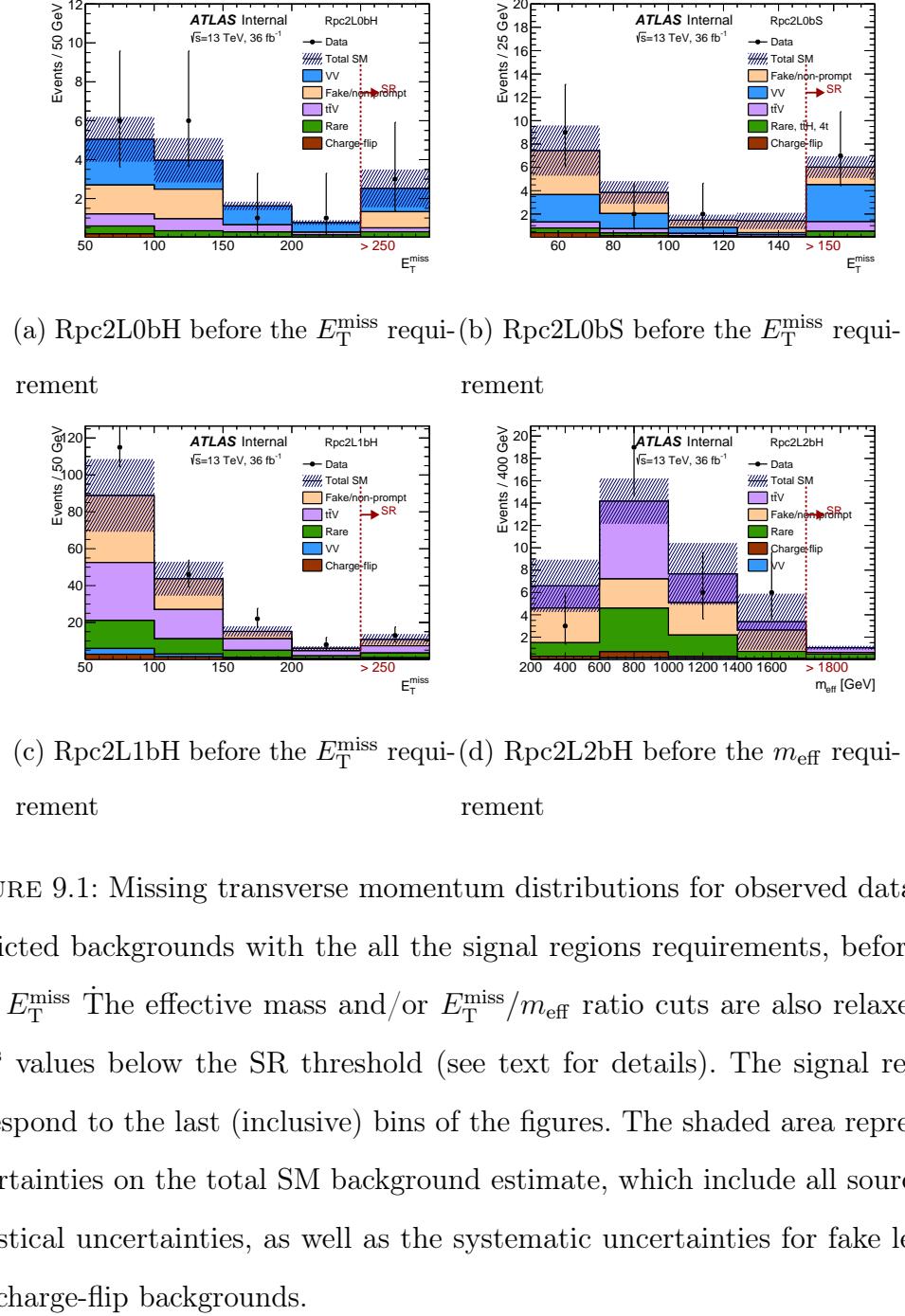


FIGURE 9.1: Missing transverse momentum distributions for observed data and predicted backgrounds with all the signal regions requirements, before the final E_T^{miss} . The effective mass and/or $E_T^{\text{miss}}/m_{\text{eff}}$ ratio cuts are also relaxed for E_T^{miss} values below the SR threshold (see text for details). The signal regions correspond to the last (inclusive) bins of the figures. The shaded area represents uncertainties on the total SM background estimate, which include all sources of statistical uncertainties, as well as the systematic uncertainties for fake lepton and charge-flip backgrounds.

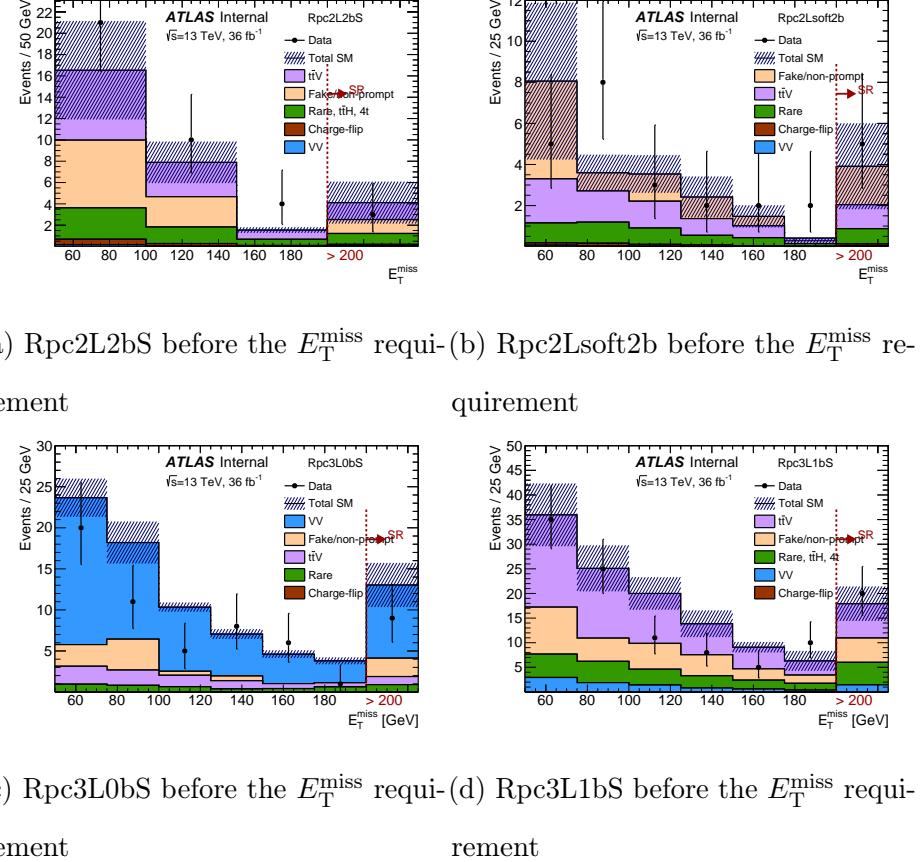


FIGURE 9.2: Missing transverse momentum distributions for observed data and predicted backgrounds with the all the signal regions requirements, before the final E_T^{miss} . The effective mass and/or $E_T^{\text{miss}}/m_{\text{eff}}$ ratio cuts are also relaxed for E_T^{miss} values below the SR threshold (see text for details). The signal regions correspond to the last (inclusive) bins of the figures. The shaded area represents uncertainties on the total SM background estimate, which include all sources of statistical uncertainties, as well as the systematic uncertainties for fake lepton and charge-flip backgrounds.

in the rare category are dominated by triboson, tWZ and $t\bar{t}WW$ production. The triboson processes generally dominate in the SRs with no b -jets, while tWZ and $t\bar{t}WW$ dominate in the SRs with one and two b -jets, respectively. Contributions from WH , ZH , tZ and $t\bar{t}t$ production never represent more than 20% of the rare background.

Figure 9.3b summarizes the contributions from the different sources of systematic uncertainty to the total SM background predictions in the signal regions. The uncertainties amount to 25–45% of the total background depending on the signal region, dominated by systematic uncertainties coming from the reducible background or the theory. The breakdown of the systematic uncertainties for each signal region is given in Table 9.2

9.2 Statistical Interpretation

In the absence of any significant deviation from the standard model predictions, the results will be interpreted to establish 95% confidence intervals using the CL_s prescription [93] based on a profile-likelihood-ratio test [81] described in Chapter 8. The interpretation will include upper limits on possible beyond the standard model contributions to the signal regions in a model independent way, as well as exclusion limits on the masses of SUSY particles in the benchmark scenarios of Figure 5.1.

Signal Region	Rpc2L2bS	Rpc2L2bH	Rpc2Lsoft1b	Rpc2Lsoft2b	Rpc2L0bS	Rpc2L0bH
$t\bar{t}W, t\bar{t}Z\gamma^*$	1.6 ± 0.4	0.44 ± 0.14	1.3 ± 0.4	1.21 ± 0.33	0.82 ± 0.31	0.20 ± 0.10
$t\bar{t}H$	0.43 ± 0.25	0.10 ± 0.06	0.45 ± 0.24	0.36 ± 0.21	0.27 ± 0.15	0.08 ± 0.07
$4t$	0.26 ± 0.13	0.18 ± 0.09	0.09 ± 0.05	0.21 ± 0.11	0.01 ± 0.01	0.02 ± 0.02
Diboson	0.10 ± 0.10	0.04 ± 0.02	0.17 ± 0.09	0.05 ± 0.03	3.1 ± 1.4	1.0 ± 0.5
Rare	0.33 ± 0.18	0.15 ± 0.09	0.18 ± 0.10	0.17 ± 0.10	0.19 ± 0.11	0.17 ± 0.10
Fake/non-prompt leptons	0.5 ± 0.6	0.15 ± 0.15	3.5 ± 2.4	1.7 ± 1.5	1.6 ± 1.0	0.9 ± 0.9
Charge-flip	0.10 ± 0.01	0.02 ± 0.01	0.08 ± 0.02	0.08 ± 0.02	0.05 ± 0.01	0.01 ± 0.01
Total Background	3.3 ± 1.0	1.08 ± 0.32	5.8 ± 2.5	3.8 ± 1.6	6.0 ± 1.8	2.4 ± 1.0
Observed	3	0	4	5	7	3

Signal Region	Rpc3L0bS	Rpc3L0bH	Rpc3L1bS	Rpc3L1bH	Rpc2L1bS	Rpc2L1bH	Rpc3LSS1b
$t\bar{t}W, t\bar{t}Z\gamma^*$	0.98 ± 0.25	0.18 ± 0.08	7.1 ± 1.1	1.54 ± 0.28	4.0 ± 1.0	4.0 ± 0.9	–
$t\bar{t}H$	0.12 ± 0.08	0.03 ± 0.02	1.4 ± 0.7	0.25 ± 0.14	1.3 ± 0.7	1.0 ± 0.6	0.22 ± 0.12
$4t$	0.02 ± 0.01	0.01 ± 0.01	0.7 ± 0.4	0.28 ± 0.15	0.34 ± 0.17	0.54 ± 0.28	–
Diboson	8.9 ± 2.9	2.6 ± 0.8	1.4 ± 0.5	0.48 ± 0.17	0.5 ± 0.3	0.7 ± 0.3	–
Rare	0.7 ± 0.4	0.29 ± 0.16	2.5 ± 1.3	0.9 ± 0.5	0.9 ± 0.5	1.0 ± 0.6	0.12 ± 0.07
Fake/non-prompt leptons	0.23 ± 0.23	0.15 ± 0.15	4.2 ± 3.1	0.5 ± 0.5	2.5 ± 2.2	2.3 ± 1.9	0.9 ± 0.7
Charge-flip	–	–	–	–	0.25 ± 0.04	0.25 ± 0.05	0.39 ± 0.08
Total Background	11.0 ± 3.0	3.3 ± 0.8	17 ± 4	3.9 ± 0.9	9.8 ± 2.9	9.8 ± 2.6	1.6 ± 0.8
Observed	9	3	20	4	14	13	1
S_{obs}^{95}	8.3	5.4	14.7	6.1	13.7	12.4	3.9
S_{exp}^{95}	$9.3^{+3.1}_{-2.3}$	$5.5^{+2.2}_{-1.5}$	$12.6^{+5.1}_{-3.4}$	$5.9^{+2.2}_{-1.8}$	$10.0^{+3.7}_{-2.6}$	$9.7^{+3.4}_{-2.6}$	$4.0^{+1.8}_{-0.3}$
$\sigma_{\text{vis}} [\text{fb}]$	0.23	0.15	0.41	0.17	0.38	0.34	0.11
$p_0 (Z)$	0.72 (–)	0.85 (–)	0.32 (0.5σ)	0.46 (0.1σ)	0.17 (1.0σ)	0.21 (0.8σ)	0.56 (–)

TABLE 9.1: Numbers of events observed in the signal regions compared with the expected backgrounds. The rare category is defined in the text. Background categories with yields shown as a “–” do not contribute to a given region (e.g. charge flips in three-lepton regions) or their estimates are below 0.01. The 95% confidence level (CL) upper limits are shown on the observed and expected numbers of BSM events, S_{obs}^{95} and S_{exp}^{95} (as well as the $\pm 1\sigma$ excursions from the expected limit), respectively. The 95% CL upper limits on the visible cross-section (σ_{vis}) are also given. Finally the p -values (p_0) give the probabilities of the observations being consistent with the estimated backgrounds. The number of equivalent Gaussian standard deviations (Z) is also shown when $p_0 < 0.5$.

Signal Region	Rpc2L2bS	Rpc2L2bH	Rpc2Lsoft1b	Rpc2Lsoft2b	Rpc2L0bS	Rpc2L0bH
Total background expectation	3.35	1.08	5.78	3.80	6.02	2.35
Total statistical	10.56%	15.67%	16.93%	15.61%	9.08%	20.87%
Total background systematic	30.41%	29.97%	43.10%	41.79%	30.51%	42.39%
Fake/non-prompt	15.46%	0.00%	38.39%	35.75%	13.46%	32.31%
Charge-flip	0.06%	0.00%	0.35%	0.53%	0.17%	0.00%
Jet Energy Scale	15.19%	11.37%	5.27%	9.28%	17.28%	8.11%
Other Jet Unc.	2.09%	2.71%	0.80%	0.99%	2.31%	3.42%
Flavor Tagging	6.27%	5.55%	0.81%	3.96%	3.33%	3.27%
Electrons	1.20%	1.72%	0.51%	0.51%	0.76%	0.74%
Muons	0.90%	1.39%	0.35%	0.51%	0.83%	0.93%
Missing transverse momentum	2.24%	1.68%	0.85%	1.50%	0.65%	0.54%
Diboson Th. Unc.	1.07%	1.39%	1.07%	0.50%	17.68%	13.54%
ttV Th. Unc.	7.33%	8.86%	5.01%	4.48%	4.06%	2.44%
Rare Th. Unc.	15.18%	19.67%	6.28%	9.75%	3.89%	5.87%
PDF	0.00%	0.00%	0.00%	0.00%	0.00%	0.00%

Signal Region	Rpc3L0bS	Rpc3L0bH	Rpc3L1bS	Rpc3L1bH	Rpc2L1bS	Rpc2L1bH	Rpc3LSS1b
Total background expectation	11.02	3.31	17.33	3.90	9.88	9.75	1.62
Total statistical	2.57%	6.05%	7.66%	7.70%	9.59%	6.65%	9.15%
Total background systematic	27.37%	25.40%	24.22%	24.02%	29.19%	26.52%	46.79%
Fake/non-prompt	1.63%	0.00%	16.50%	9.73%	19.93%	18.05%	44.45%
Charge-flip	0.00%	0.00%	0.00%	0.00%	0.40%	0.41%	4.32%
Jet Energy Scale	9.78%	8.98%	5.54%	4.20%	11.71%	10.40%	0.02%
Other Jet Unc.	3.41%	2.55%	0.70%	2.30%	1.42%	1.46%	0.20%
Flavor Tagging	2.79%	2.93%	2.22%	2.82%	1.32%	1.38%	0.32%
Electrons	1.78%	2.16%	1.66%	2.47%	0.67%	0.89%	0.41%
Muons	1.73%	2.12%	1.25%	1.79%	0.80%	0.92%	0.41%
Missing transverse momentum	0.78%	0.53%	0.38%	0.59%	1.70%	1.06%	0.00%
Diboson Th. Unc.	24.28%	21.58%	2.57%	3.78%	1.87%	2.50%	0.00%
ttV Th. Unc.	1.49%	1.76%	5.34%	5.56%	6.96%	5.72%	0.00%
Rare Th. Unc.	4.02%	5.02%	13.19%	18.11%	12.68%	13.16%	10.49%
PDF	0.00%	0.00%	0.00%	0.00%	0.00%	0.00%	0.00%

TABLE 9.2: Breakdown of the dominant systematic uncertainties on background estimates in the various signal regions. Note that the individual uncertainties can be correlated, and do not necessarily add up quadratically to the total background uncertainty. The percentages show the size of the uncertainty relative to the total expected background.

9.2.1 Model independent discovery and upper limits

Table 9.1 presents 95% confidence level (CL) observed (expected) model-independent upper limits on the number of BSM events, S_{obs}^{95} (S_{exp}^{95}), that may contribute to the signal regions. Normalizing these by the integrated luminosity L of the data sample, they can be interpreted as upper limits on the visible BSM cross-section (σ_{vis}), defined as $\sigma_{\text{vis}} = \sigma_{\text{prod}} \times A \times \epsilon = S_{\text{obs}}^{95}/L$, where σ_{prod} is the production cross-section, A the acceptance and ϵ the reconstruction efficiency. The largest deviation of the data from the background prediction corresponds to an excess of 1.0 standard deviation in the Rpc2L1bS SR.

9.2.2 Model dependent exclusion limits

Exclusion limits at 95% CL are set on the masses of the superpartners involved in the SUSY benchmark scenarios considered. Apart from the NUHM2 model, simplified models are used, corresponding to a single production mode and with 100% branching ratio to a specific decay chain, with the masses of the SUSY particles not involved in the process set to very high values.

In order to determine which signal region is used to set an exclusion limit on a particular model, the expected CL_s value is computed for each signal region at a given point in the signal parameter space. The signal region with the smallest expected CL_s value (more disagreement with data under the signal hypothesis) is used to set an exclusion limit on the model. An example on what signal region is performing best at a given model using the decay is shown in Figure 9.4.

Figures 9.6 and 9.8 show the exclusion limits in all the models considered in Figure 5.1 and the NUHM2 model. The assumptions about the decay chain considered for the different SUSY particles are stated above each figure. For each region of the signal parameter space, the SR with the best expected sensitivity is chosen.

Each one of the Figures 9.6 contain :

- Observed limit (thick solid red line) : all uncertainties are included in the fit as nuisance parameters, with the exception of the theoretical signal uncertainties (PDF, scales) on the inclusive cross section.
- Expected limit (less thick long-dashed black line) : all uncertainties are included in the fit as nuisance parameters, with the exception of the theoretical signal uncertainties (PDF, scales) on the inclusive cross section.
- $\pm 1\sigma$ lines around the observed limit (thin dark-red dotted) : re-run limit calculation while increasing or decreasing the signal cross section by the theoretical signal uncertainties (PDF, scales).
- $\pm 1\sigma$ band around expected limit (yellow band) : represents the $\pm 1\sigma$ uncertainty from the fit.

The limits set are compared with the existing limits set by other ATLAS SUSY searches [96, 97]. For the models shown in Figure 9.6, the mass limits on gluinos and bottom squarks are up to 400GeV higher than the previous limits, reflecting the improvements in the signal region definitions as well as the increase in integrated luminosity. Gluinos with masses up to 1.75 TeV are excluded in scenarios with a light $\tilde{\chi}_1^0$ in Figure 9.5a. This limit is extended to 1.87 TeV when

$\tilde{\chi}_2^0$ and slepton masses are in between the gluino and the $\tilde{\chi}_1^0$ masses (Figure 9.6b).

More generally, gluino masses below 1.57 TeV and bottom squarks with masses below 700 GeV are excluded in models with a massless LSP. The “compressed” regions, where SUSY particle masses are close to each other, are also better covered and LSP masses up to 1200 and 250 GeV are excluded in the gluino and bottom squark pair-production models, respectively. Of particular interest is the observed exclusion of models producing gluino pairs with an off-shell top quark in the decay (Figure 5.1b), see Figure 9.5a. In this case, models are excluded for mass differences between the gluino and neutralino of 205 GeV (only 35 GeV larger than the minimum mass difference for decays into two on-shell W bosons and two b -quarks) for a gluino mass below 0.9 TeV. The Rpc3LSS1b SR allows the exclusion of top squarks with masses below 700 GeV when the top squark decays to a top quark and a cascade of electroweakinos $\tilde{\chi}_2^0 \rightarrow \tilde{\chi}_1^\pm W^\mp \rightarrow W^* W^\mp \tilde{\chi}_1^0$ (see Figure 9.7b for the conditions on the sparticle masses).

Finally, in the NUHM2 model with low fine-tuning, values of the parameter $m_{1/2}$ below 615 GeV are excluded, corresponding to gluino masses below 1500 GeV (Figure 9.8).

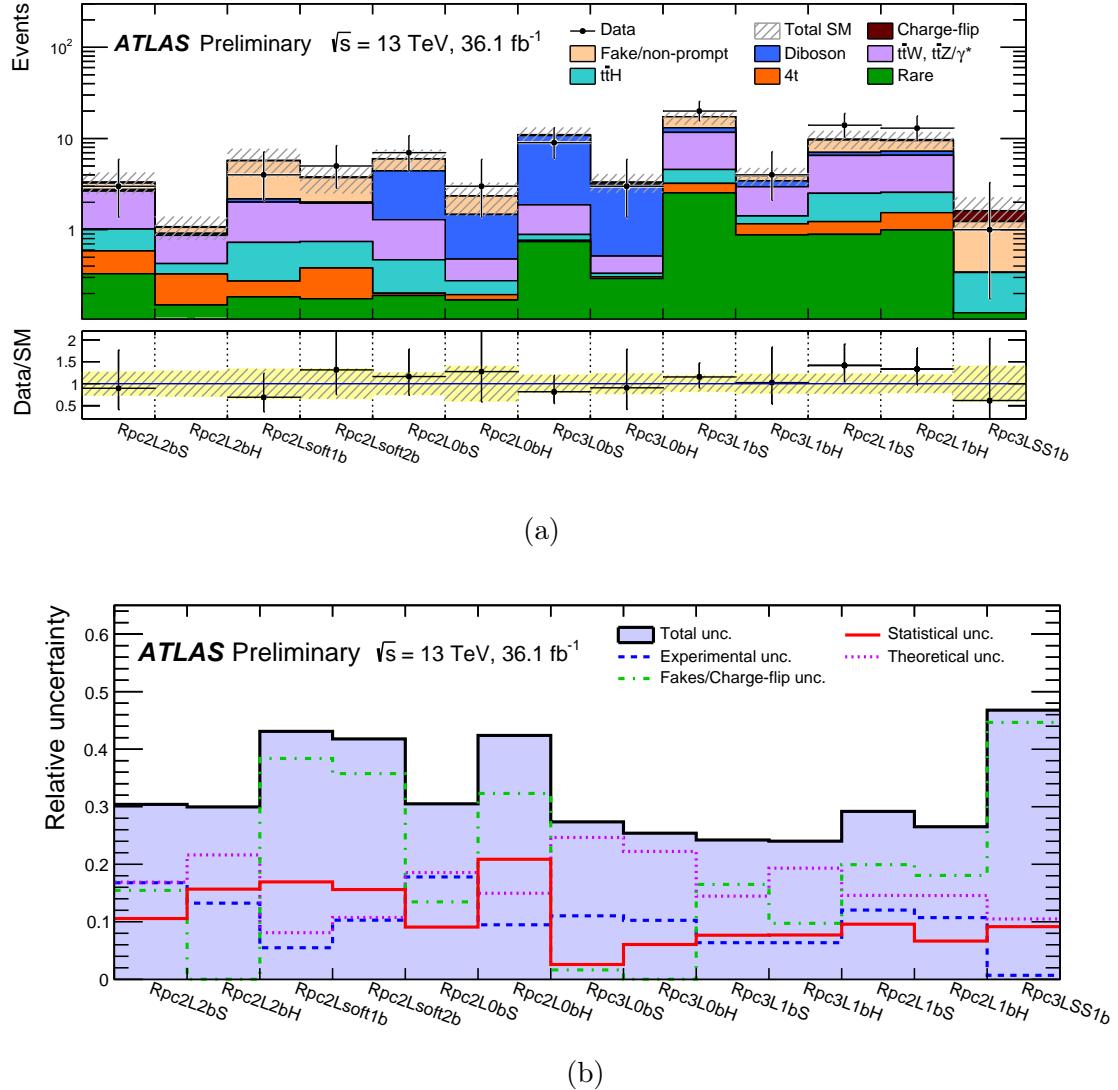


FIGURE 9.3: Comparison of (a) the observed and expected event yields in each signal region and (b) the relative uncertainties in the total background yield estimate. For the latter, “statistical uncertainty” corresponds to reducible and irreducible background statistical uncertainties. The background predictions correspond to those presented in Table 9.1 and the rare category is explained in the text.

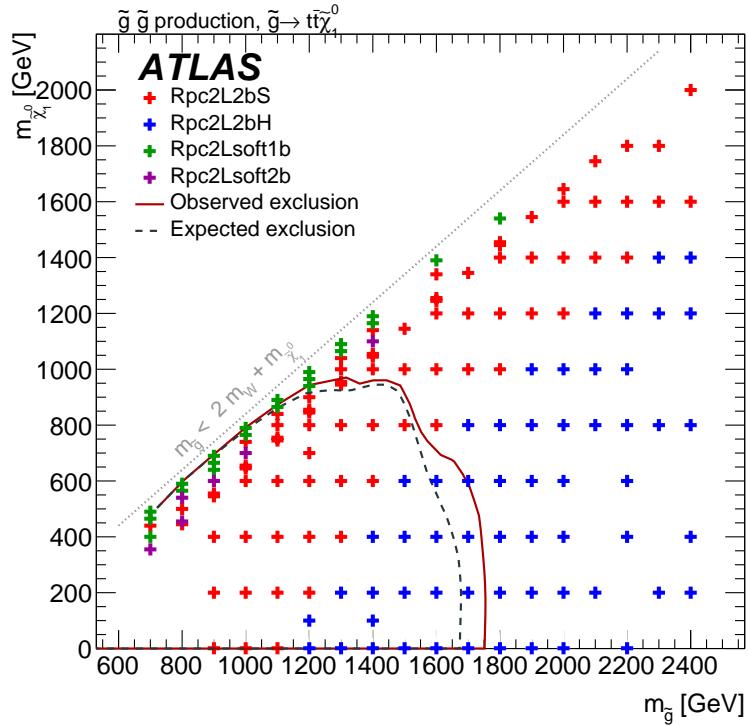


FIGURE 9.4: Illustration of the best expected signal region per signal grid point for the $\tilde{g} \rightarrow t\bar{t}\tilde{\chi}_1^0$ (Fig. 5.1a) model. This mapping is used for the final combined exclusion limits.

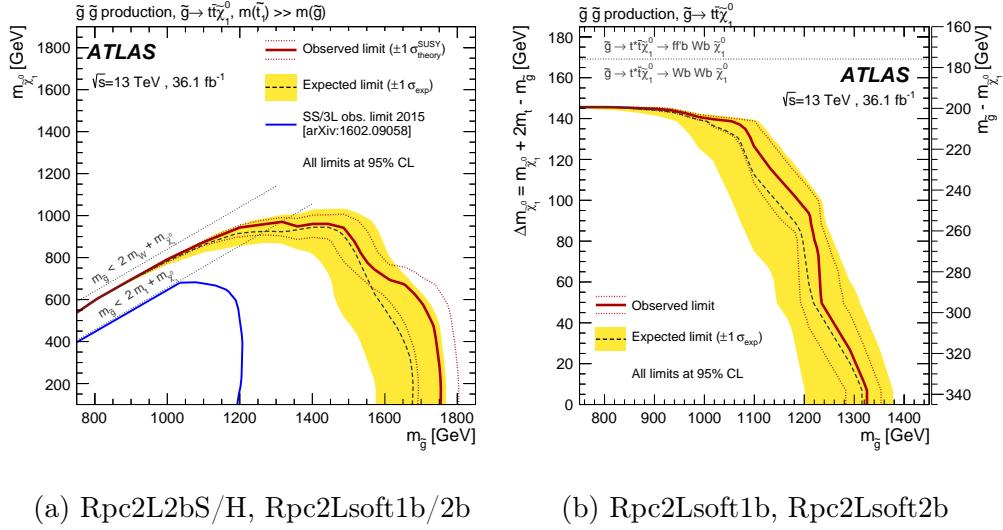


FIGURE 9.5: Observed and expected exclusion limits on the \tilde{g} and $\tilde{\chi}_1^0$ masses for (a) the model in Figure 5.1a and (b) the model in Figure 5.1b. Figure (b) is a zoomed version of Figure (a) in the mass-parameter space where there is at least one top-quark off-shell decay. All limits are computed at 95% CL. The dotted lines around the observed limit illustrate the change in the observed limit as the nominal signal cross-section is scaled up and down by the theoretical uncertainty. The contours of the band around the expected limit are the $\pm 1\sigma$ results, including all uncertainties except the theoretical ones in the signal cross-section.

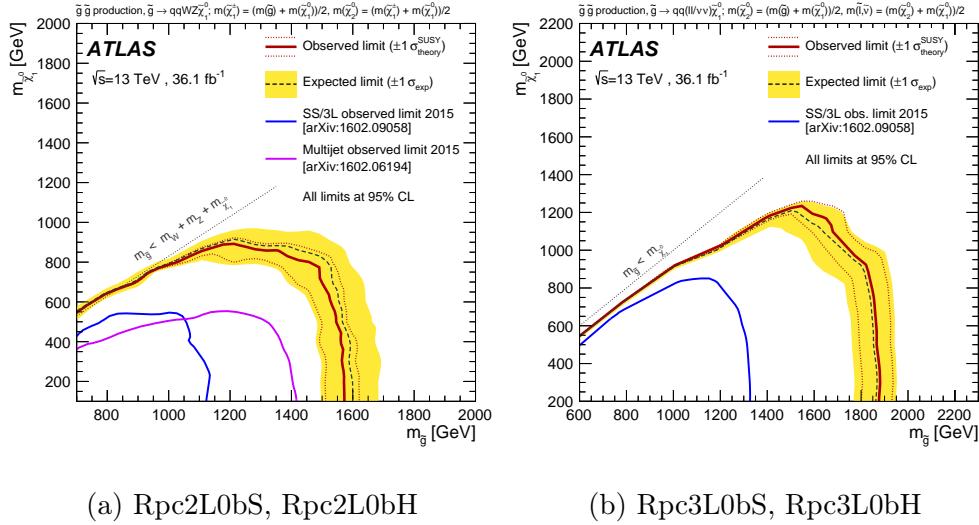


FIGURE 9.6: Observed and expected exclusion limits on the \tilde{g} and $\tilde{\chi}_1^0$ masses for (a) the model in Figure 5.1c and (b) the model in Figure 5.1d. All limits are computed at 95% CL. The dotted lines around the observed limit illustrate the change in the observed limit as the nominal signal cross-section is scaled up and down by the theoretical uncertainty. The contours of the band around the expected limit are the $\pm 1\sigma$ results, including all uncertainties except the theoretical ones in the signal cross-section.

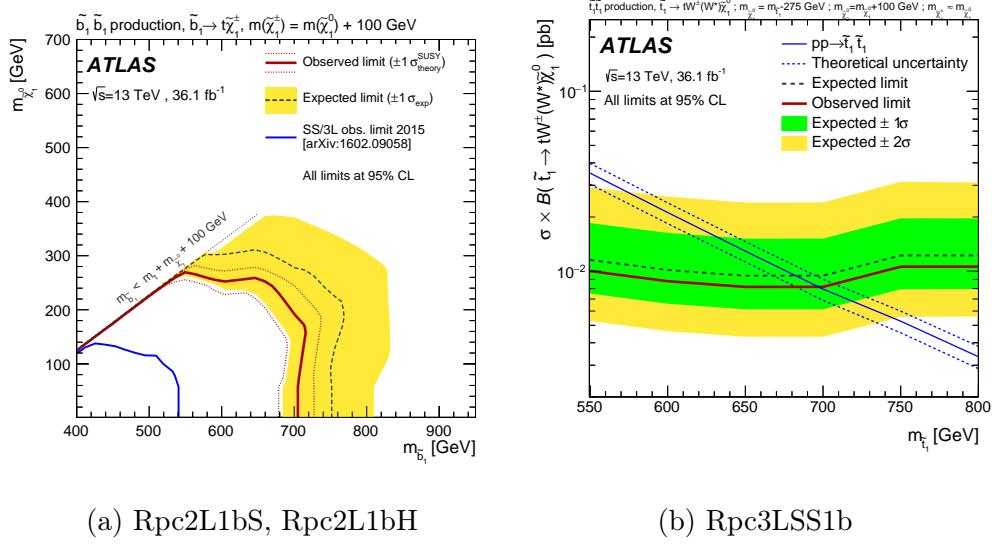


FIGURE 9.7: Observed and expected exclusion limits on the \tilde{b}_1 , \tilde{t}_1 , and $\tilde{\chi}_1^0$ masses for (a) the model in Figure 5.1e and (b) the model in Figure 5.1f. The two models are complementary where the one-dimensional change in stop masses (b) is along the grayed diagonal in the sbottom plot (b). All limits are computed at 95% CL. The dotted lines around the observed limit illustrate the change in the observed limit as the nominal signal cross-section is scaled up and down by the theoretical uncertainty. The contours of the band around the expected limit are the $\pm 1\sigma$ results ($\pm 2\sigma$ is also considered in (b)), including all uncertainties except the theoretical ones in the signal cross-section.

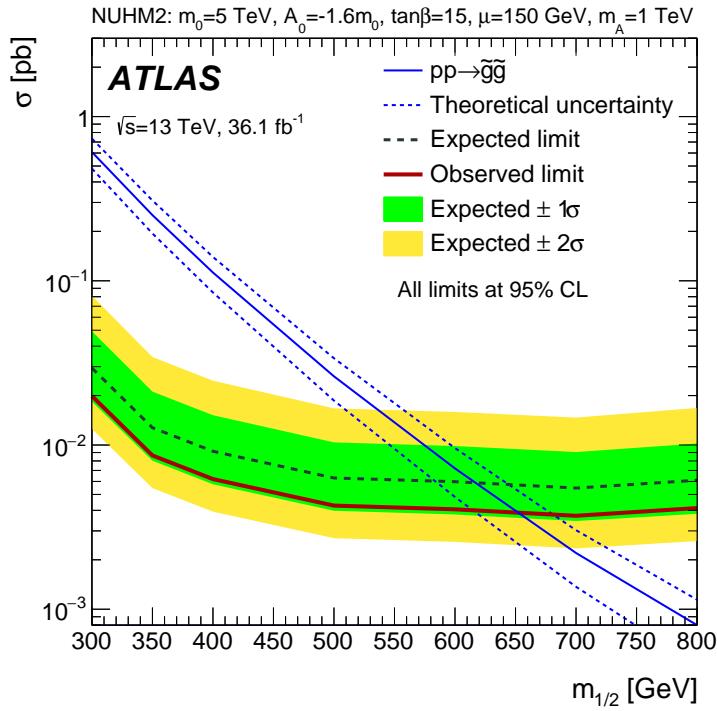


FIGURE 9.8: Observed and expected exclusion limits as a function of $m_{1/2}$ in the NUHM2 model [94, 95]. The signal region Rpc2L2bH is used to obtain the limits. The contours of the green (yellow) band around the expected limit are the $\pm 1\sigma$ ($\pm 2\sigma$) results, including all uncertainties. The limits are computed at 95% CL.

Chapitre 10

Conclusions

A search for supersymmetry in events with two same-sign leptons or at least three leptons, multiple jets, b -jets and large E_T^{miss} and/or large m_{eff} is presented. The analysis is performed with proton–proton collision data at $\sqrt{s} = 13\text{TeV}$ collected in 2015 and 2016 with the ATLAS detector at the Large Hadron Collider corresponding to an integrated luminosity of 36.1 fb^{-1} . With no significant excess over the Standard Model prediction observed, results are interpreted in the framework of simplified models featuring gluino and squark production in R -parity-conserving and R -parity-violating scenarios. Lower limits on particle masses are derived at 95% confidence level. In the $\tilde{g}\tilde{g}$ simplified RPC models considered, gluinos with masses up to 1.87 TeV are excluded in scenarios with a light $\tilde{\chi}_1^0$. RPC models with bottom squark masses below 700 GeV are also excluded in a $\tilde{b}_1\tilde{b}_1^*$ simplified model with $\tilde{b}_1 \rightarrow tW^-\tilde{\chi}_1^0$ and a light $\tilde{\chi}_1^0$. In RPV scenarios, masses of down squark-rights are probed up to $m_{\tilde{d}_{\text{R}}} \approx 500\text{ GeV}$. All models with gluino masses below 1.3 TeV are excluded, greatly extending the previous exclusion limits. Model-independent limits on the cross-section of a possible signal contribution to the signal regions are set.

Annexe A

Auxiliary material

A.1 Signal region with best exclusion

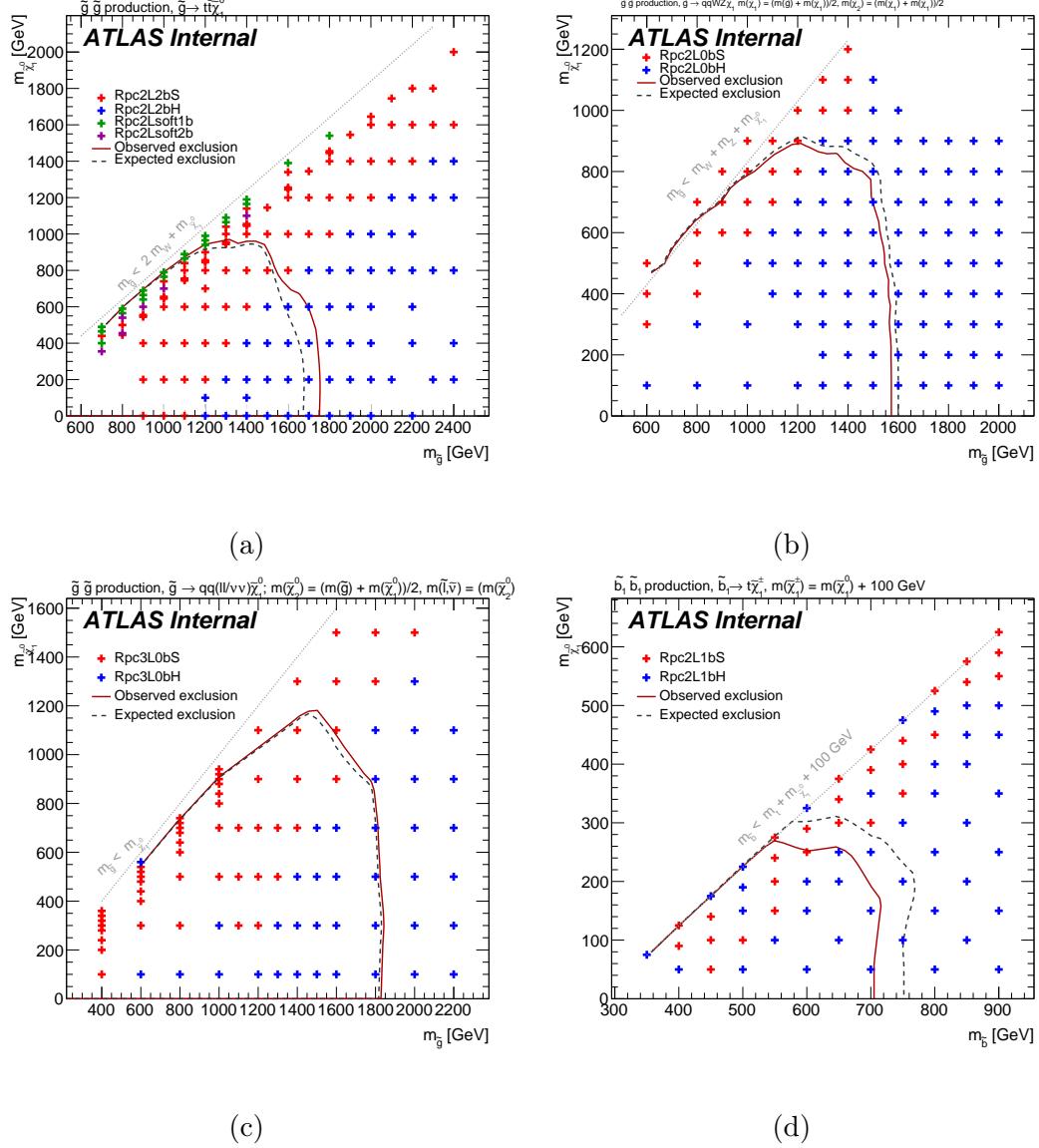


FIGURE A.1: Illustration of the best expected signal region per signal grid point for the $\tilde{g} \rightarrow q\bar{q}(\ell\ell/\nu\nu)\tilde{\chi}_1^0$ (Fig. 5.1a), $\tilde{g} \rightarrow q\bar{q}'WZ\tilde{\chi}_1^0$ (Fig. 5.1c), $\tilde{g} \rightarrow q\bar{q}'\ell/\nu\ell/\nu\tilde{\chi}_1^0$ (Fig. 5.1d), $\tilde{b} \rightarrow tW\tilde{\chi}_1^0$ (Fig. 5.1e) models. This mapping is used for the final combined exclusion limits.

A.2 Upper limit on cross section

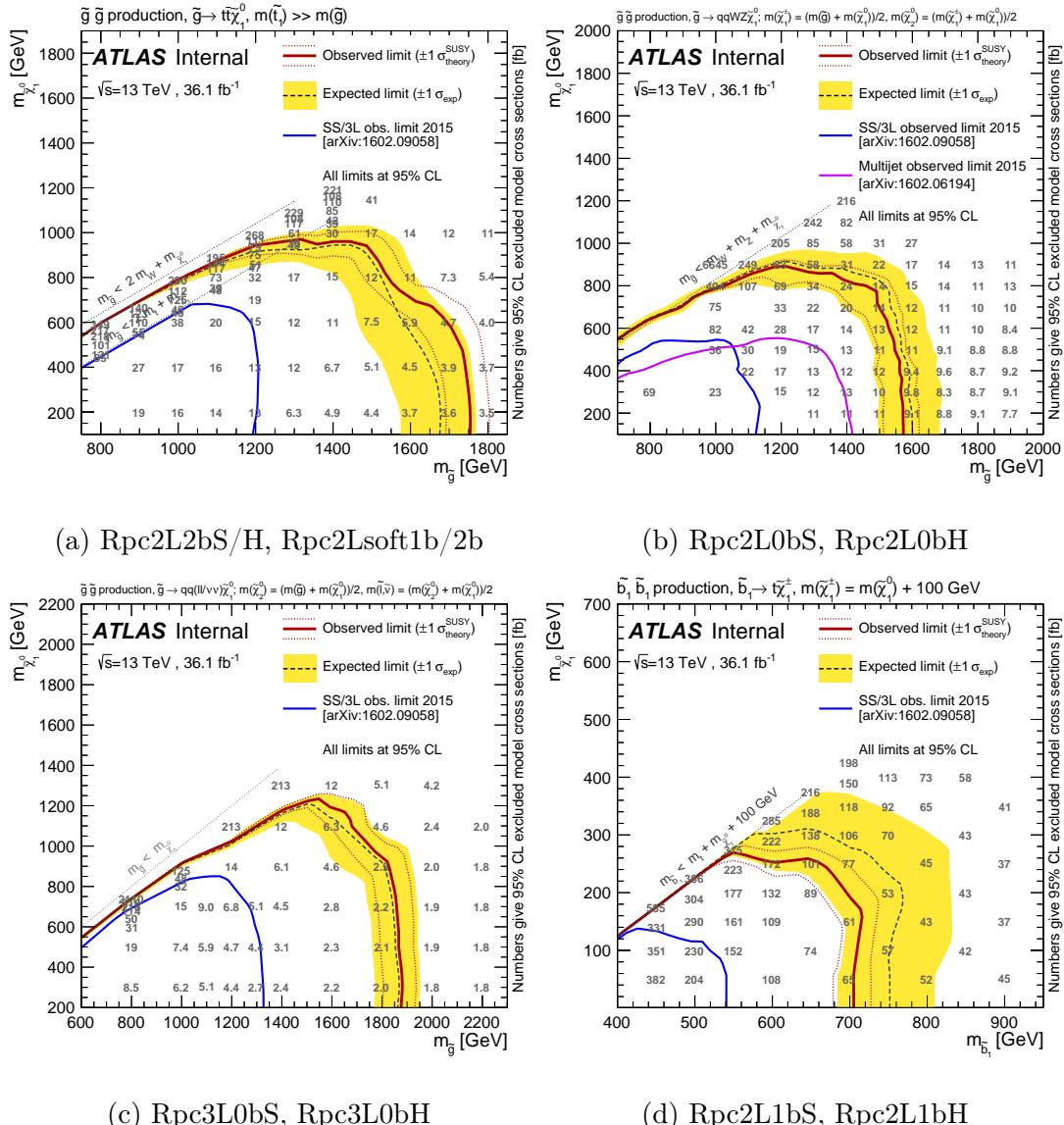


FIGURE A.2: Observed and expected exclusion limits on the \tilde{g} , \tilde{b}_1 , \tilde{t}_1 and $\tilde{\chi}_1^0$ masses in the context of RPC SUSY scenarios with simplified mass spectra. The signal regions used to obtain the limits are specified in the subtitle of each scenario. All limits are computed at 95% CL. The dotted lines around the observed limit illustrate the change in the observed limit as the nominal signal cross-section is scaled up and down by the theoretical uncertainty. The contours of the band around the expected limit are the $\pm 1\sigma$ results, including all uncertainties except the theoretical ones on the signal cross-section. In Figures A.2a–A.2d, the grey diagonal lines indicate the kinematic limit for the decays in each specified scenario and results are compared with the observed limits obtained by previous ATLAS searches [96, 97]. The grey numbers show 95% CL upper limits on production cross-sections (in fb) obtained using the signal efficiency and acceptance specific

A.3 Signal region cutflow

Rpc2L2bS, $\tilde{g}\tilde{g}$ production, $\tilde{g} \rightarrow t\bar{t}\tilde{\chi}_1^0$ $m_{\tilde{g}} = 1.5\text{TeV}$, $m_{\tilde{\chi}_1^0} = 800\text{GeV}$	
MC events generated	98000
Expected for 36.1 fb^{-1}	5.1×10^2
≥ 2 SS leptons ($p_T > 20\text{GeV}$)	19.96 ± 0.35
Trigger	19.17 ± 0.35
≥ 2 b -jets ($p_T > 20\text{GeV}$)	16.10 ± 0.32
≥ 6 jets ($p_T > 25\text{GeV}$)	13.11 ± 0.28
$E_T^{\text{miss}} > 200\text{GeV}$	10.17 ± 0.26
$m_{\text{eff}} > 0.6\text{TeV}$	10.17 ± 0.26
$E_T^{\text{miss}} > 0.25 \times m_{\text{eff}}$	5.94 ± 0.20

TABLE A.1: Number of signal events at different stages of the Rpc2L2bS signal region selection. Only statistical uncertainties are shown.

A.4 Acceptance and Efficiency

Rpc2L2bH, $\tilde{g}\tilde{g}$ production, $\tilde{g} \rightarrow t\bar{t}\tilde{\chi}_1^0$ $m_{\tilde{g}} = 1.7\text{TeV}$, $m_{\tilde{\chi}_1^0} = 200\text{GeV}$	
MC events generated	98000
Expected for 36.1 fb^{-1}	1.7×10^2
≥ 2 SS leptons ($p_T > 20\text{GeV}$)	7.32 ± 0.13
Trigger	7.19 ± 0.13
≥ 2 <i>b</i> -jets ($p_T > 20\text{GeV}$)	5.81 ± 0.11
≥ 6 jets ($p_T > 40\text{GeV}$)	4.92 ± 0.11
$m_{\text{eff}} > 1.8\text{TeV}$	3.93 ± 0.09
$E_T^{\text{miss}} > 0.15 \times m_{\text{eff}}$	3.12 ± 0.08

TABLE A.2: Number of signal events at different stages of the Rpc2L2bH signal region selection. Only statistical uncertainties are shown.

Rpc2Lsoft1b, $\tilde{g}\tilde{g}$ production, $\tilde{g} \rightarrow tWb\tilde{\chi}_1^0$ $m_{\tilde{g}} = 1.2\text{TeV}$, $m_{\tilde{\chi}_1^0} = 940\text{GeV}$	
MC events generated	50000
Expected for 36.1 fb^{-1}	3.1×10^3
≥ 2 SS leptons ($100 > p_T > 20, 10 \text{ GeV}$)	101.9 ± 2.7
Trigger	89.3 ± 2.5
≥ 1 b -jet ($p_T > 20\text{GeV}$)	75.1 ± 2.3
≥ 6 jets ($p_T > 25\text{GeV}$)	31.5 ± 1.5
$E_T^{\text{miss}} > 100\text{GeV}$	23.0 ± 1.3
$E_T^{\text{miss}} > 0.3 \times m_{\text{eff}}$	6.5 ± 0.7

TABLE A.3: Number of signal events at different stages of the Rpc2Lsoft1b signal region selection. Only statistical uncertainties are shown.

Rpc2Lsoft2b, $\tilde{g}\tilde{g}$ production, $\tilde{g} \rightarrow tWb\tilde{\chi}_1^0$ $m_{\tilde{g}} = 1.2\text{TeV}$, $m_{\tilde{\chi}_1^0} = 900\text{GeV}$	
MC events generated	50000
Expected for 36.1 fb^{-1}	3.1×10^3
≥ 2 SS leptons ($100 > p_T > 20, 10 \text{ GeV}$)	91.8 ± 2.6
Trigger	79.7 ± 2.4
≥ 2 b -jets ($p_T > 20\text{GeV}$)	41.3 ± 1.7
≥ 6 jets ($p_T > 25\text{GeV}$)	21.4 ± 1.2
$E_T^{\text{miss}} > 200\text{GeV}$	8.7 ± 0.7
$m_{\text{eff}} > 0.6\text{TeV}$	8.7 ± 0.7
$E_T^{\text{miss}} > 0.25 \times m_{\text{eff}}$	6.7 ± 0.6

TABLE A.4: Number of signal events at different stages of the Rpc2Lsoft2b signal region selection. Only statistical uncertainties are shown.

Rpc2L0bS, $\tilde{g}\tilde{g}$ production, $\tilde{g} \rightarrow q\bar{q}'WZ\tilde{\chi}_1^0$ $m_{\tilde{g}} = 1.2\text{TeV}$, $(m_{\tilde{\chi}_1^\pm} - 150) = (m_{\tilde{\chi}_2^0} - 75) = m_{\tilde{\chi}_1^0} = 900\text{GeV}$	
MC events generated	19000
Expected for 36.1 fb^{-1}	3.1×10^3
≥ 2 SS leptons ($p_T > 20\text{GeV}$)	64 ± 4
Trigger	58.6 ± 3.3
no b -jet ($p_T > 20\text{GeV}$)	46.3 ± 3.0
≥ 6 jets ($p_T > 25\text{GeV}$)	26.6 ± 2.4
$E_T^{\text{miss}} > 150\text{GeV}$	16.3 ± 2.0
$E_T^{\text{miss}} > 0.25 \times m_{\text{eff}}$	9.0 ± 1.3

TABLE A.5: Number of signal events at different stages of the Rpc2L0bS signal region selection. Only statistical uncertainties are shown.

Rpc2L0bH, $\tilde{g}\tilde{g}$ production, $\tilde{g} \rightarrow q\bar{q}'WZ\tilde{\chi}_1^0$ $m_{\tilde{g}} = 1.6\text{TeV}$, $(m_{\tilde{\chi}_1^\pm} - 750) = (m_{\tilde{\chi}_2^0} - 375) = m_{\tilde{\chi}_1^0} = 100\text{GeV}$	
MC events generated	20000
Expected for 36.1 fb^{-1}	2.9×10^2
≥ 2 SS leptons ($p_T > 20\text{GeV}$)	12.8 ± 0.5
Trigger	12.5 ± 0.5
no b -jet ($p_T > 20\text{GeV}$)	8.5 ± 0.4
≥ 6 jets ($p_T > 40\text{GeV}$)	7.12 ± 0.35
$E_T^{\text{miss}} > 250\text{GeV}$	5.13 ± 0.29
$m_{\text{eff}} > 0.9\text{TeV}$	5.13 ± 0.29

TABLE A.6: Number of signal events at different stages of the Rpc2L0bH signal region selection. Only statistical uncertainties are shown.

Rpc3L0bS, $\tilde{g}\tilde{g}$ production, $\tilde{g} \rightarrow q\bar{q}(\tilde{\ell}\ell/\tilde{\nu}\nu)$ $m_{\tilde{g}} = 1.4\text{TeV}$, $(m_{\tilde{\chi}_2^0} - 150) = (m_{\tilde{\ell},\tilde{\nu}} - 75) = m_{\tilde{\chi}_1^0} = 1100\text{GeV}$	
MC events generated	20000
Expected for 36.1 fb^{-1}	9.1×10^2
≥ 3 leptons ($p_T > 20, 20, 10$ GeV)	76.9 ± 2.1
Trigger	76.0 ± 2.0
no b -jet ($p_T > 20\text{GeV}$)	67.5 ± 1.9
≥ 4 jets ($p_T > 40\text{GeV}$)	31.6 ± 1.3
$E_T^{\text{miss}} > 200\text{GeV}$	17.1 ± 1.0
$m_{\text{eff}} > 0.6\text{TeV}$	17.1 ± 1.0

TABLE A.7: Number of signal events at different stages of the Rpc3L0bS signal region selection. Only statistical uncertainties are shown.

Rpc3L0bH, $\tilde{g}\tilde{g}$ production, $\tilde{g} \rightarrow q\bar{q}(\tilde{\ell}\ell/\tilde{\nu}\nu)$ $m_{\tilde{g}} = 1.8\text{TeV}$, $(m_{\tilde{\chi}_2^0} - 850) = (m_{\tilde{\ell},\tilde{\nu}} - 375) = m_{\tilde{\chi}_1^0} = 100\text{GeV}$	
MC events generated	20000
Expected for 36.1 fb^{-1}	1.0×10^2
≥ 3 leptons ($p_T > 20, 20, 10$ GeV)	9.98 ± 0.25
Trigger	9.94 ± 0.25
no b -jet ($p_T > 20\text{GeV}$)	8.44 ± 0.23
≥ 4 jets ($p_T > 40\text{GeV}$)	7.79 ± 0.22
$E_T^{\text{miss}} > 200\text{GeV}$	6.58 ± 0.21
$m_{\text{eff}} > 1.6\text{TeV}$	6.56 ± 0.21

TABLE A.8: Number of signal events at different stages of the Rpc3L0bH signal region selection. Only statistical uncertainties are shown.

Rpc2L1bS, $\tilde{b}_1 \tilde{b}_1^*$ production, $\tilde{b}_1 \rightarrow t\tilde{\chi}_1^- \rightarrow tW^-\tilde{\chi}_1^0$ $m_{\tilde{b}_1} = 600\text{GeV}$, $m_{\tilde{\chi}_1^\pm} = 350\text{GeV}$, $m_{\tilde{\chi}_1^0} = 250\text{GeV}$	
MC events generated	10000
Expected for 36.1 fb^{-1}	6.3×10^3
≥ 2 SS leptons ($p_T > 20\text{GeV}$)	221 ± 4
Trigger	201 ± 4
≥ 1 b -jet ($p_T > 20\text{GeV}$)	173 ± 4
≥ 6 jets ($p_T > 25\text{GeV}$)	66.3 ± 2.2
$E_T^{\text{miss}} > 150\text{GeV}$	36.5 ± 1.7
$m_{\text{eff}} > 0.6\text{TeV}$	36.1 ± 1.7
$E_T^{\text{miss}} > 0.25 \times m_{\text{eff}}$	15.1 ± 1.1

TABLE A.9: Number of signal events at different stages of the Rpc2L1bS signal region selection. Only statistical uncertainties are shown.

Rpc2L1bH, $\tilde{b}_1 \tilde{b}_1^*$ production, $\tilde{b}_1 \rightarrow t \tilde{\chi}_1^-$	
$m_{\tilde{b}_1} = 750\text{GeV}$, $m_{\tilde{\chi}_1^\pm} = 200\text{GeV}$, $m_{\tilde{\chi}_1^0} = 100\text{GeV}$	
MC events generated	10000
Expected for 36.1 fb^{-1}	1.6×10^3
≥ 2 SS leptons ($p_T > 20\text{GeV}$)	71.1 ± 1.2
Trigger	66.4 ± 1.2
≥ 1 b -jet ($p_T > 20\text{GeV}$)	56.6 ± 1.1
≥ 6 jets ($p_T > 25\text{GeV}$)	27.7 ± 0.7
$E_T^{\text{miss}} > 250\text{GeV}$	12.5 ± 0.5
$E_T^{\text{miss}} > 0.2 \times m_{\text{eff}}$	9.5 ± 0.4

TABLE A.10: Number of signal events at different stages of the Rpc2L1bH signal region selection. Only statistical uncertainties are shown.

Rpc3LSS1b, $\tilde{t}_1 \tilde{t}_1^*$ production, $\tilde{t}_1 \rightarrow t \tilde{\chi}_2^0 \rightarrow \bar{t} W^\pm \tilde{\chi}_1^\mp$	
$m_{\tilde{t}_1} = 700\text{GeV}$, $m_{\tilde{\chi}_2^0} = 525\text{GeV}$, $m_{\tilde{\chi}_1^\pm} \approx m_{\tilde{\chi}_1^0} = 425\text{GeV}$	
MC events generated	5000
Expected for 36.1 fb^{-1}	2.4×10^3
≥ 3 SS leptons ($p_T > 20, 20, 10 \text{ GeV}$), $Z \rightarrow e^\pm e^\pm$ veto	4.6 ± 0.5
Trigger	4.5 ± 0.5
≥ 1 b -jet ($p_T > 20\text{GeV}$)	3.6 ± 0.4

TABLE A.11: Number of signal events at different stages of the Rpc3LSS1b signal region selection. Only statistical uncertainties are shown.

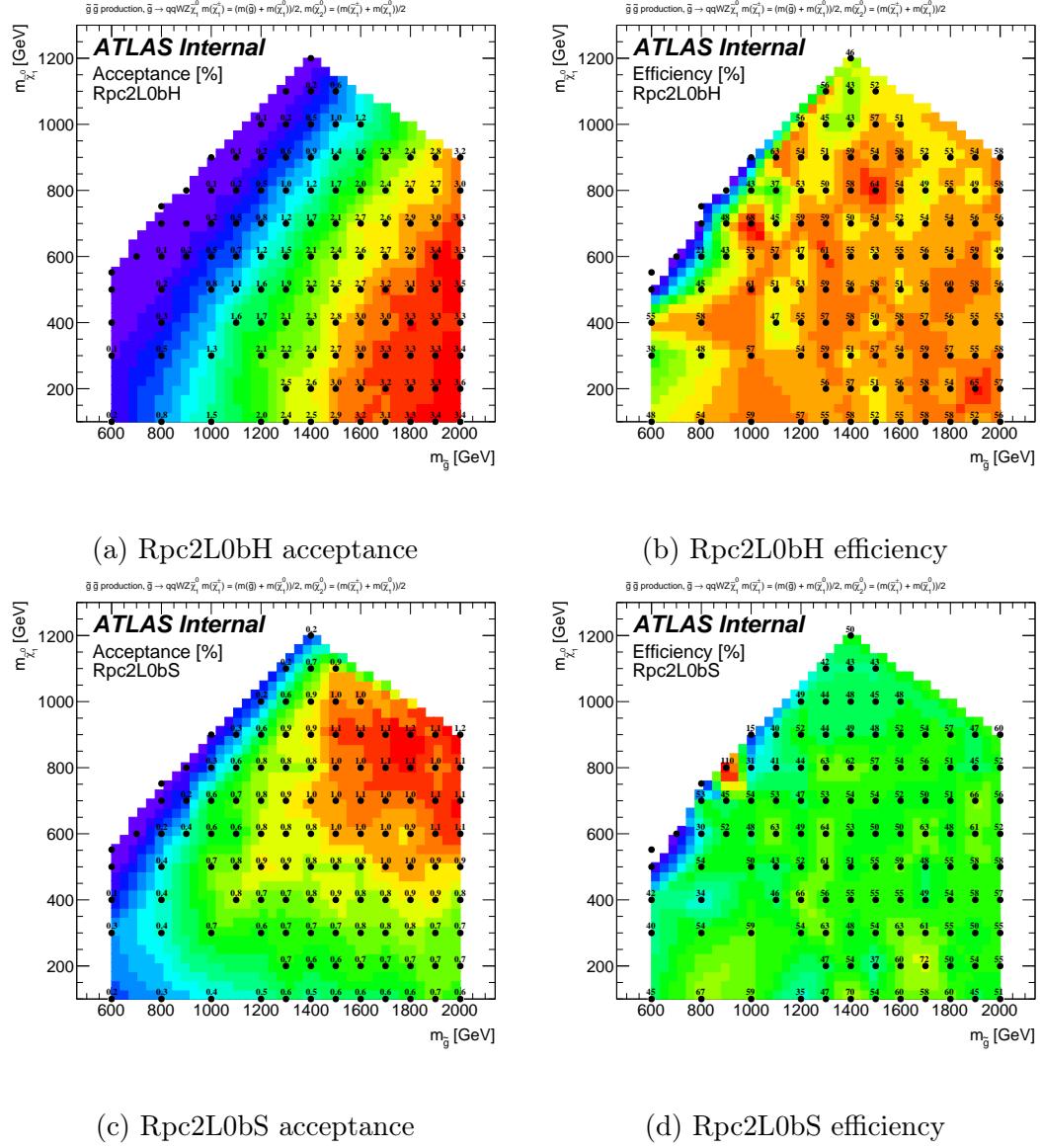


FIGURE A.3: Signal acceptance (a,c) and reconstruction efficiency (b,d) for simplified models of $\tilde{g}\tilde{g}$ production with $\tilde{g} \rightarrow q\bar{q}'WZ\tilde{\chi}_1^0$ decays, in the signal regions Rpc2L0bH (a,b) and Rpc2L0bS (c,d).

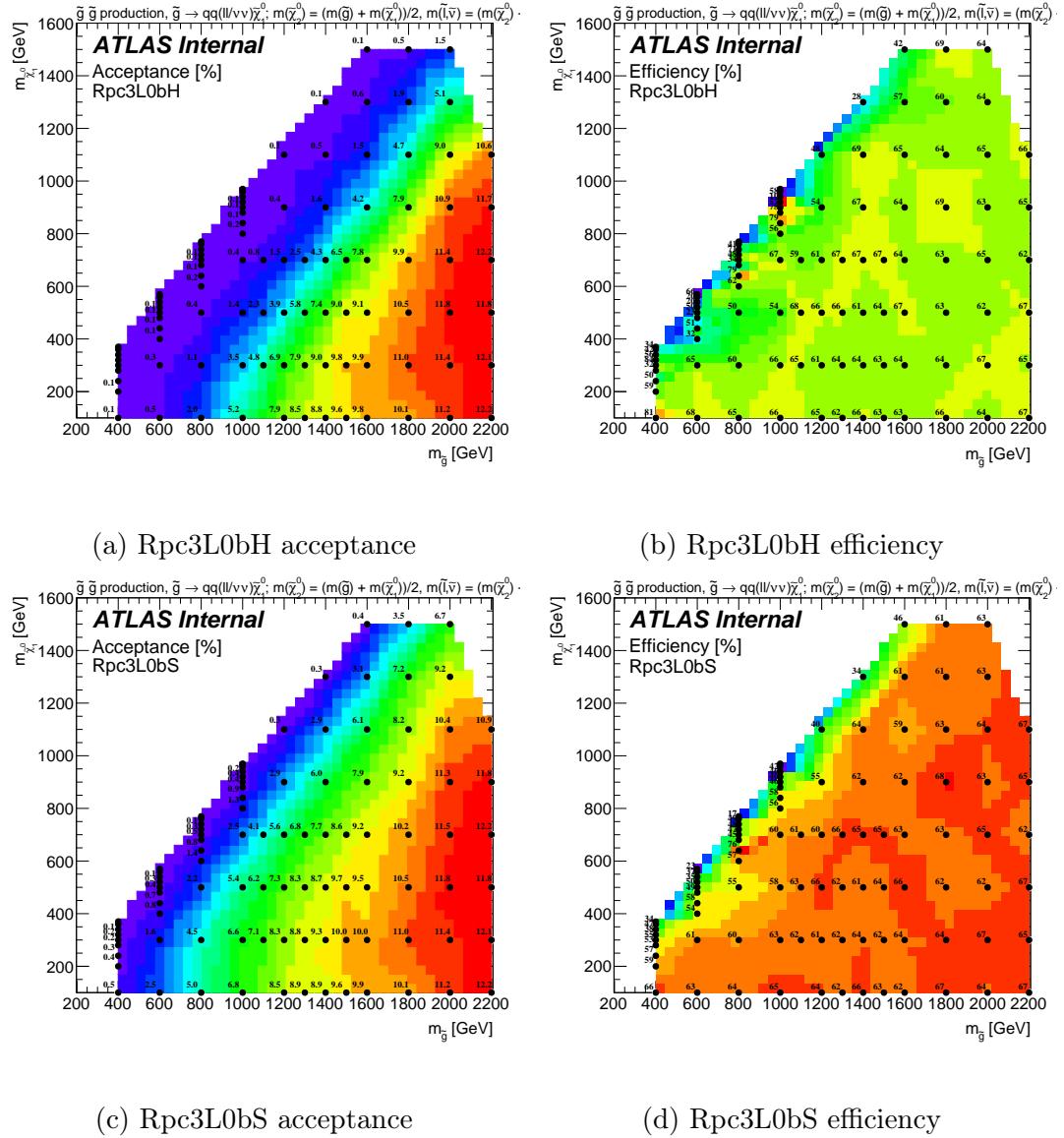


FIGURE A.4: Signal acceptance (a,c) and reconstruction efficiency (b,d) for simplified models of $\tilde{g}\tilde{g}$ production with $\tilde{g} \rightarrow q\bar{q}(\ell\ell/\nu\nu)\tilde{\chi}_1^0$ decays, in the signal regions Rpc3L0bH (a,b) and Rpc3L0bS (c,d).

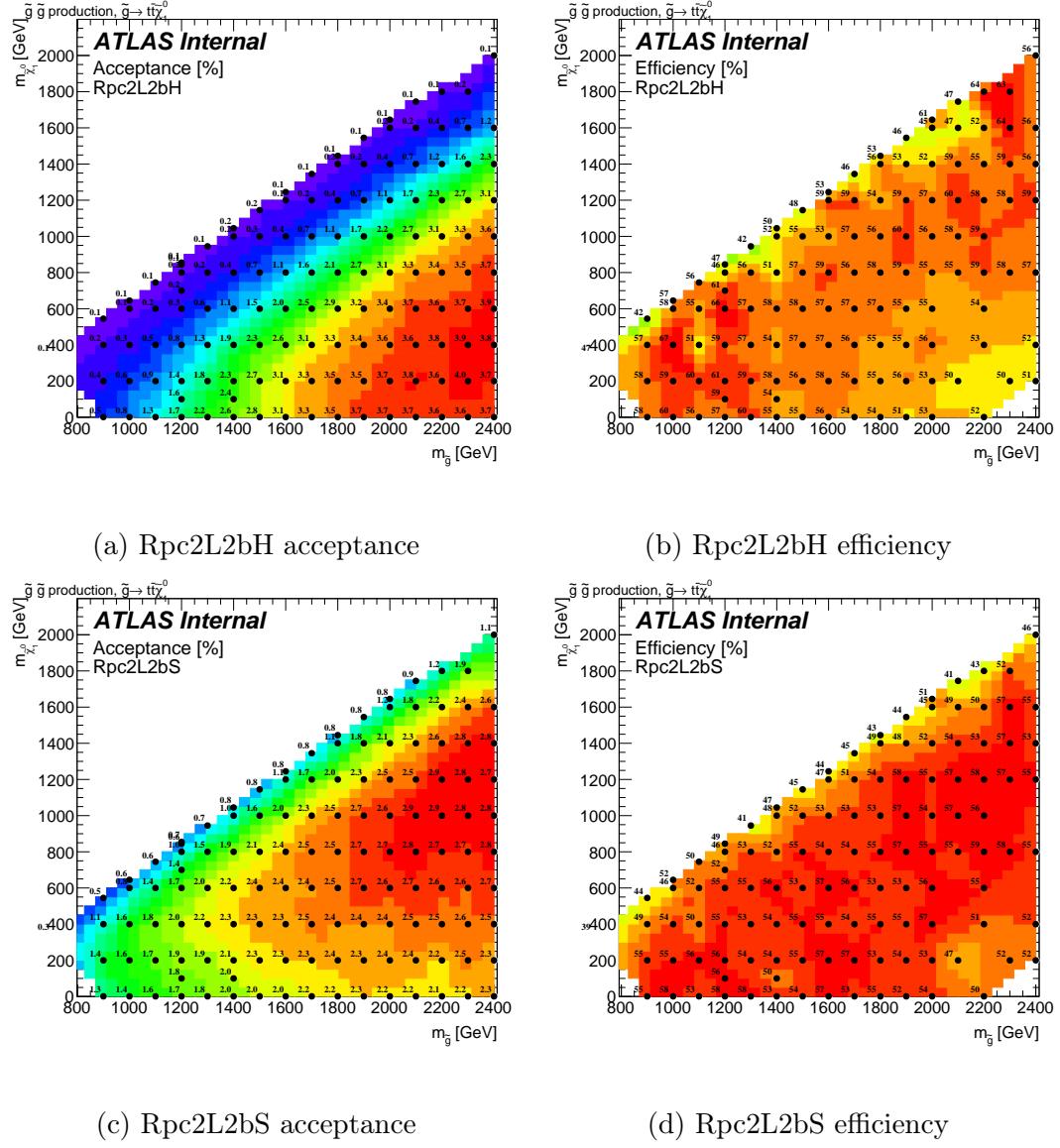


FIGURE A.5: Signal acceptance (a,c) and reconstruction efficiency (b,d) for simplified models of $\tilde{g}\tilde{g}$ production with $\tilde{g} \rightarrow t\bar{t}\tilde{\chi}_1^0$ decays, in the signal regions Rpc2L2bH (a,b) and Rpc2L2bS (c,d).

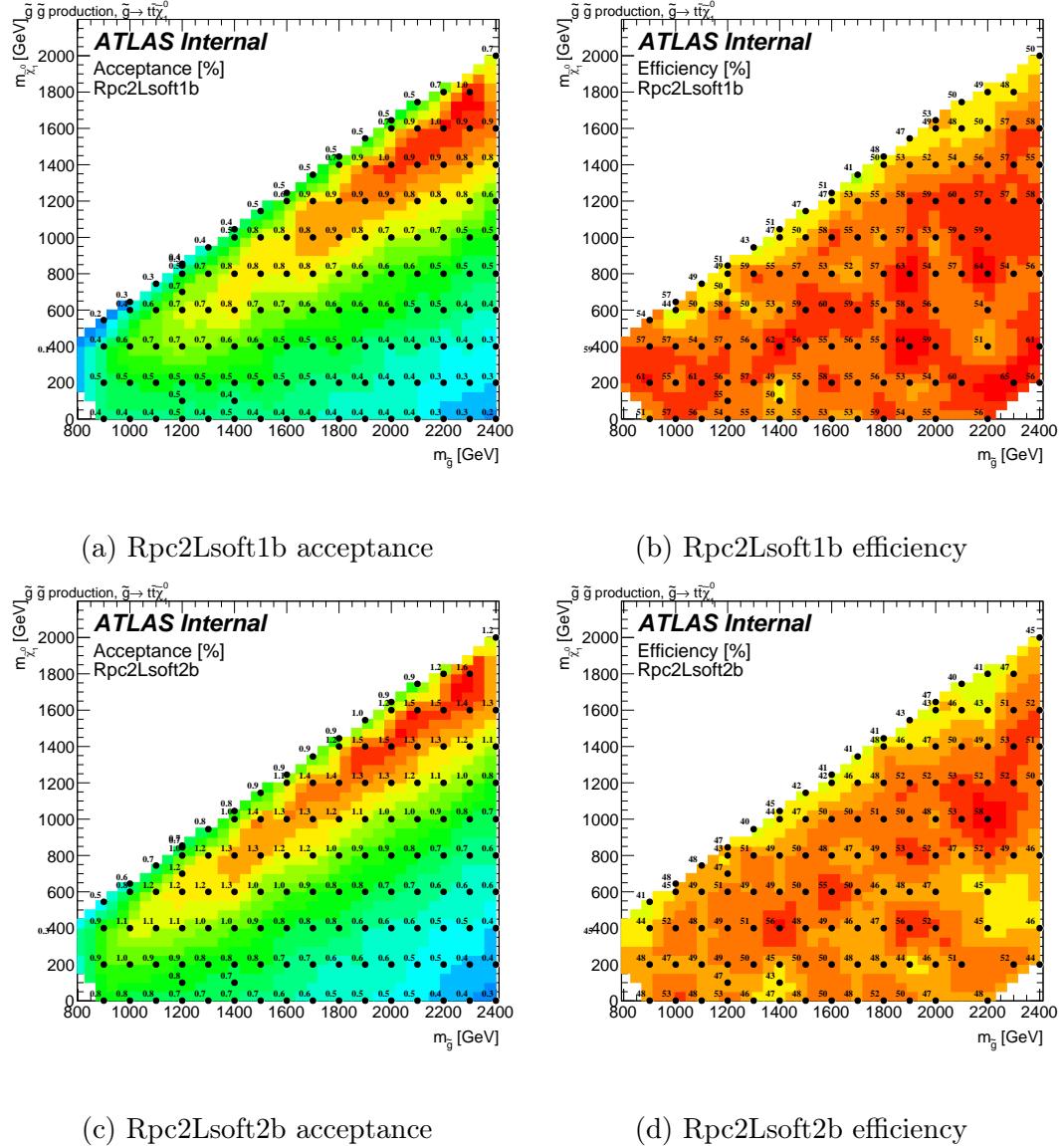


FIGURE A.6: Signal acceptance (a,c) and reconstruction efficiency (b,d) for simplified models of $\tilde{g}\tilde{g}$ production with $\tilde{g} \rightarrow t\bar{t}\tilde{\chi}_1^0$ decays, in the signal regions Rpc2Lsoft1b (a,b) and Rpc2Lsoft2b (c,d).

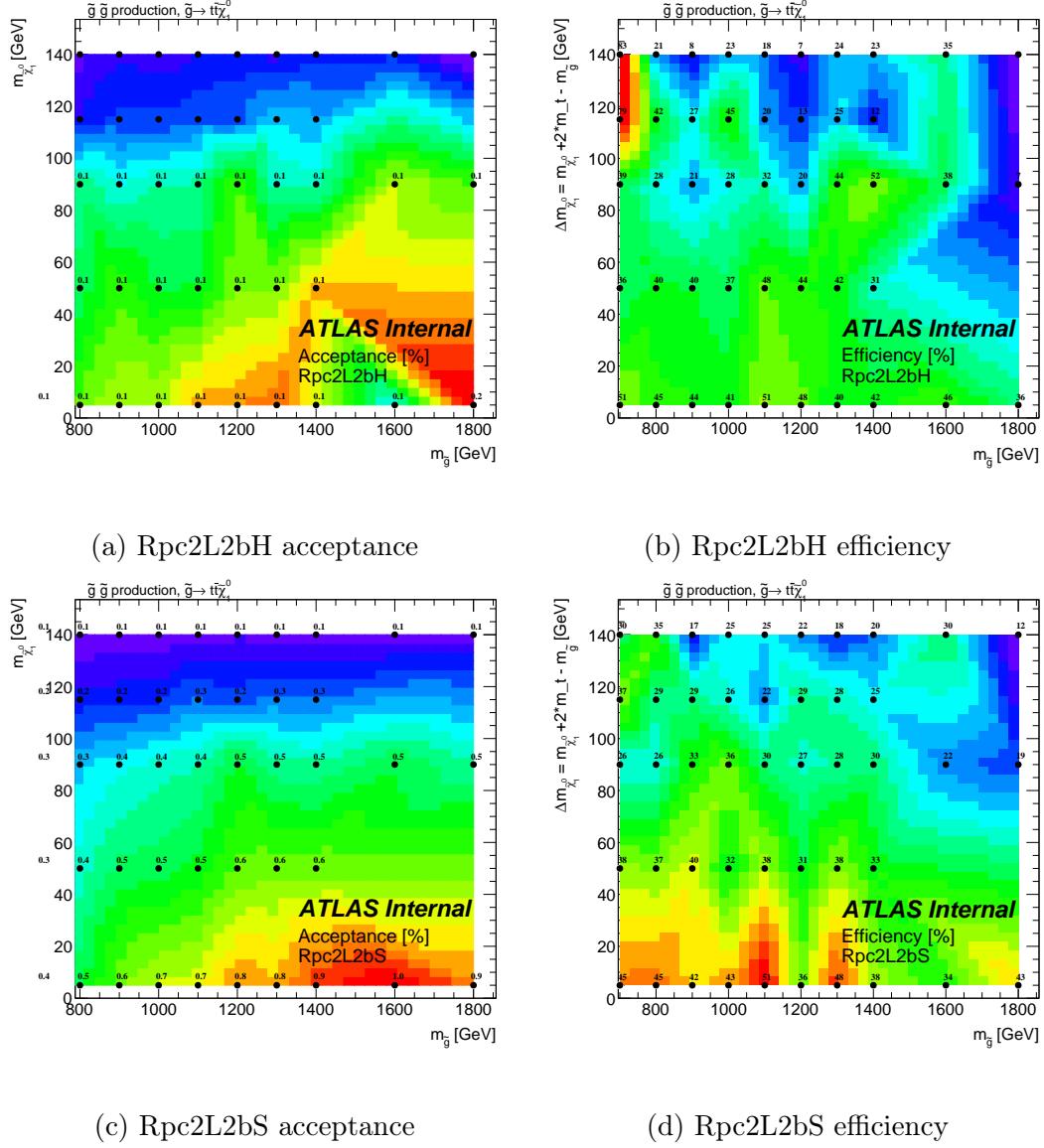


FIGURE A.7: Signal acceptance (a,c) and reconstruction efficiency (b,d) for simplified models of $\tilde{g}\tilde{g}$ production with $\tilde{g} \rightarrow tWb\tilde{\chi}_1^0$ decays (region with $\Delta m(\tilde{g}, \tilde{\chi}_1^0) < 2m_t$), in the signal regions Rpc2L2bH (a,b) and Rpc2L2bS (c,d).

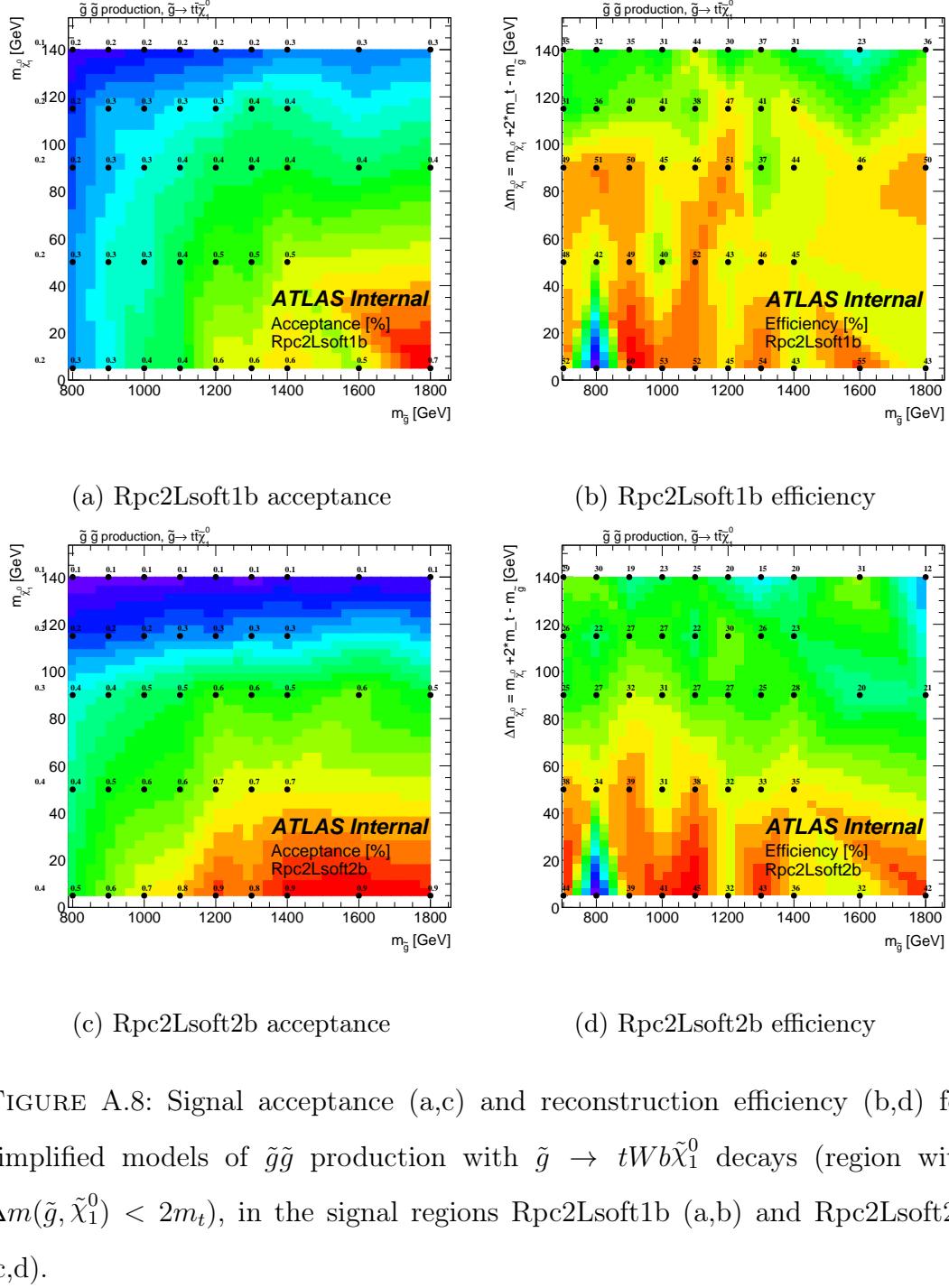


FIGURE A.8: Signal acceptance (a,c) and reconstruction efficiency (b,d) for simplified models of $\tilde{g}\tilde{g}$ production with $\tilde{g} \rightarrow tWb\tilde{\chi}_1^0$ decays (region with $\Delta m(\tilde{g}, \tilde{\chi}_1^0) < 2m_t$), in the signal regions Rpc2Lsoft1b (a,b) and Rpc2Lsoft2b (c,d).

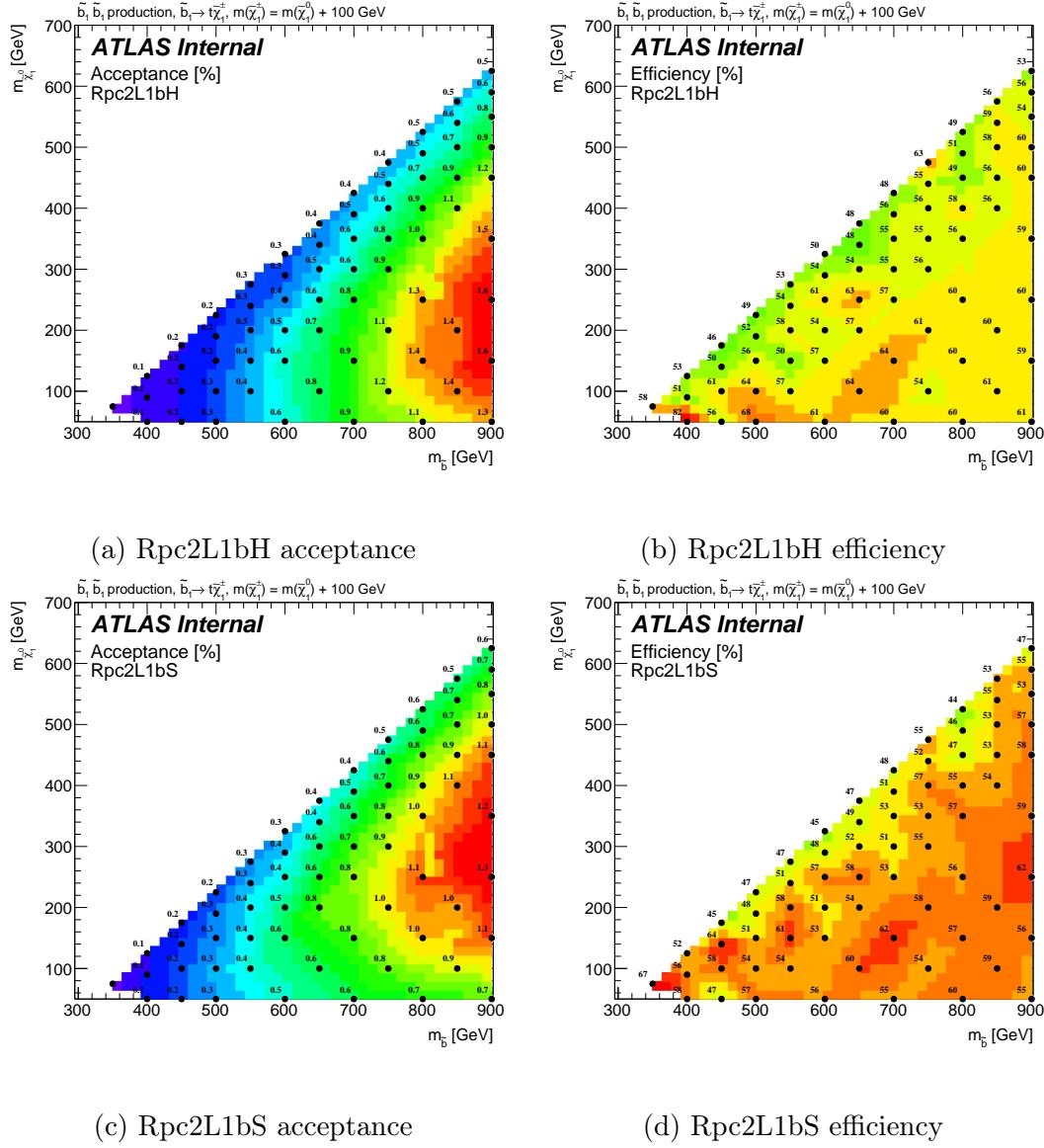


FIGURE A.9: Signal acceptance (a,c) and reconstruction efficiency (b,d) for simplified models of $\tilde{b}_1 \tilde{b}_1^*$ production with $\tilde{b}_1 \rightarrow tW^- \tilde{\chi}_1^0$ decays, in the signal regions Rpc2L1bH (a,b) and Rpc2L1bS (c,d).

Rpc2L2bH, $\tilde{g}\tilde{g}$ production in the NUHM2 model							
$m_{\tilde{g}}$ [GeV]	300	350	400	500	600	700	800
Acceptance	0.8%	1.6%	2.1%	3.2%	3.5%	4.4%	4.0%
Efficiency	43%	49%	50%	49%	48%	43%	49%

TABLE A.12: Rpc2L2bH signal region acceptance and reconstruction efficiency for $\tilde{g}\tilde{g}$ production in the NUHM2 model.

Rpc3LSS1b, $\tilde{t}_1\tilde{t}_1^*$ production, $\tilde{t}_1 \rightarrow t\tilde{\chi}_2^0, \tilde{\chi}_2^0 \rightarrow W^\mp\tilde{\chi}_1^\pm$						
$m_{\tilde{t}_1}$ [GeV]	550	600	650	700	750	800
Acceptance	0.3%	0.3%	0.3%	0.4%	0.4%	0.4%
Efficiency	36%	42%	44%	37%	33%	30%

TABLE A.13: Rpc3LSS1b signal region acceptance and reconstruction efficiency for $\tilde{t}_1\tilde{t}_1^*$ production with $\tilde{t}_1 \rightarrow t\tilde{\chi}_2^0$ ($\tilde{\chi}_2^0 \rightarrow W\tilde{\chi}_1^\pm$) decays.

Annexe B

Validation of background estimates

Annexe C

HistFitter Tables

Ex

Bibliographie

- [1] CDF Collaboration Collaboration, F. Abe et al., *Phys. Rev. Lett.* **74** (1995) 2626–2631, <https://link.aps.org/doi/10.1103/PhysRevLett.74.2626>.
- [2] D0 Collaboration, S. Abachi et al., *Phys. Rev. Lett.* **74** (1995) 2632–2637, [arXiv:hep-ex/9503003 \[hep-ex\]](https://arxiv.org/abs/hep-ex/9503003).
- [3] ATLAS Collaboration, G. Aad et al., *Phys. Lett. B* **716** (2012) 1–29, [arXiv:1207.7214 \[hep-ex\]](https://arxiv.org/abs/1207.7214).
- [4] CMS Collaboration, S. Chatrchyan et al., *Phys. Lett. B* **716** (2012) 30–61, [arXiv:1207.7235 \[hep-ex\]](https://arxiv.org/abs/1207.7235).
- [5] F. Zwicky, *Helvetica Physica Acta* **6** (1933) 110–127.
- [6] W. Hu and S. Dodelson, *Ann. Rev. Astron. Astrophys.* **40** (2002) 171–216, [arXiv:astro-ph/0110414 \[astro-ph\]](https://arxiv.org/abs/astro-ph/0110414).
- [7] A. Challinor and H. Peiris, [arXiv:0903.5158 \[astro-ph.CO\]](https://arxiv.org/abs/0903.5158).
- [8] G. Bertone, D. Hooper, and J. Silk, *Phys. Rept.* **405** (2005) 279–390, [arXiv:hep-ph/0404175 \[hep-ph\]](https://arxiv.org/abs/hep-ph/0404175).
- [9] Supernova Cosmology Project Collaboration, S. Perlmutter et al., *Astrophys. J.* **517** (1999) 565–586, [arXiv:astro-ph/9812133 \[astro-ph\]](https://arxiv.org/abs/astro-ph/9812133).
- [10] Planck Collaboration, R. Adam et al., *Astron. Astrophys.* **594** (2016) A1, [arXiv:1502.01582 \[astro-ph.CO\]](https://arxiv.org/abs/1502.01582).
- [11] Planck Collaboration, P. A. R. Ade et al., *Astron. Astrophys.* **594** (2016) A13, [arXiv:1502.01589 \[astro-ph.CO\]](https://arxiv.org/abs/1502.01589).
- [12] SDSS Collaboration, R. Scranton et al., [arXiv:astro-ph/0307335 \[astro-ph\]](https://arxiv.org/abs/astro-ph/0307335).
- [13] N. Garelli, *J. Phys. : Conf. Ser.* 513 (2014) 012007.
- [14] A. Negri, *J. Phys. Conf. Ser.* **396** (2012) 012033, <http://stacks.iop.org/1742-6596/396/i=1/a=012033>.
- [15] ATLAS Collaboration, ATLAS-TDR-12, 1998.
- [16] ATLAS Collaboration, [CERN/LHCC/2003-022](https://cds.cern.ch/record/2003-022), CERN Geneva 2003.
- [17] A. Borga et al., PoS(TIPP2014)205, 2014.
- [18] *Luminosity Public Results in Run 2*, <https://twiki.cern.ch/twiki/bin/view/AtlasPublic/LuminosityPublicResultsRun2>. Accessed : 2016-08-01.
- [19] A. Collaboration, *Technical Design Report for the Phase-I Upgrade of the ATLAS TDAQ System*, Tech. Rep. CERN-LHCC-2013-018, Sep, 2013. <https://cds.cern.ch/record/1602235>.
- [20] W. P. Vazquez et al., *Nuclear and Particle Physics Proceedings* **273 - 275** (2016) 939 – 944, <http://www.sciencedirect.com/science/article/pii/S2405601415006355>.

- [21] A. Negri et al., Journal of Physics : Conference Series **396** (2012) 012033, <http://stacks.iop.org/1742-6596/396/i=1/a=012033>.
- [22] J. Alwall, R. Frederix, S. Frixione, V. Hirschi, F. Maltoni, O. Mattelaer, H. S. Shao, T. Stelzer, P. Torrielli, and M. Zaro, *JHEP* **07** (2014) 079, [arXiv:1405.0301 \[hep-ph\]](https://arxiv.org/abs/1405.0301).
- [23] T. Sjöstrand, S. Mrenna, and P. Skands, *Comput. Phys. Commun.* **178** (2008) 852–867, [arXiv:0710.3820 \[hep-ph\]](https://arxiv.org/abs/0710.3820).
- [24] GEANT4 Collaboration, S. Agostinelli et al., *Nucl. Instrum. Meth. A* **506** (2003) 250–303.
- [25] ATLAS Collaboration, ATL-PHYS-PUB-2010-013, [http://cds.cern.ch/record/1300517](https://cds.cern.ch/record/1300517).
- [26] ATLAS Collaboration, ATLAS-CONF-2016-024, 2016, <https://cds.cern.ch/record/2157687>.
- [27] M. Cacciari, G. P. Salam, and G. Soyez, *JHEP* **0804** (2008) 063, [arXiv:0802.1189 \[hep-ph\]](https://arxiv.org/abs/0802.1189).
- [28] ATLAS Collaboration, [arXiv:1603.02934 \[hep-ex\]](https://arxiv.org/abs/1603.02934).
- [29] ATLAS Collaboration, G. Aad et al., [arXiv:1603.02934 \[hep-ex\]](https://arxiv.org/abs/1603.02934).
- [30] ATLAS Collaboration, ATL-PHYS-PUB-2015-034, 2015, <https://cds.cern.ch/record/2042098>.
- [31] R. Blair et al., 2008 JINST 3 P04001.
- [32] R. Blair et al., ATL-DQ-TR-0020, 2012.
- [33] CERN, <http://www.cerntech.hu/products/15-tilar-express.html>, October 15, 2015.
- [34] H. Engel and U. Kebschull, 2013 JINST 8 C12016.
- [35] A. Borga et al., 2015 JINST 10 C02022.
- [36] ATLAS Collaboration, *JHEP* **10** (2015) 054, [arXiv:1507.05525 \[hep-ex\]](https://arxiv.org/abs/1507.05525).
- [37] ATLAS Collaboration, *Eur. Phys. J. C* **75** (2015) 510, [arXiv:1506.08616 \[hep-ex\]](https://arxiv.org/abs/1506.08616).
- [38] ATLAS Collaboration, Atlas-conf-2015-066, <https://cds.cern.ch/record/2114833>.
- [39] P. Huang, A. Ismail, I. Low, and C. E. M. Wagner, *Phys. Rev. D* **92** (2015) 075035, [arXiv:1507.01601 \[hep-ph\]](https://arxiv.org/abs/1507.01601).
- [40] ATLAS Collaboration, *Phys. Rev. D* **88** (2013).
- [41] H. Baer, V. Barger, P. Huang, D. Mickelson, A. Mustafayev, W. Sreethawong, and X. Tata, *JHEP* **12** (2013) 013, [arXiv:1310.4858 \[hep-ph\]](https://arxiv.org/abs/1310.4858), [Erratum : JHEP06,053(2015)].
- [42] H. Baer, V. Barger, P. Huang, D. Mickelson, A. Mustafayev, W. Sreethawong, and X. Tata, *Phys. Rev. Lett.* **110** (2013) 151801, [arXiv:1302.5816 \[hep-ph\]](https://arxiv.org/abs/1302.5816).

- [43] H. Baer, V. Barger, M. Savoy, and X. Tata, Phys. Rev. **D94** (2016) 035025, arXiv:1604.07438 [hep-ph].
- [44] ATLAS Collaboration, ATLAS Collaboration, Eur. Phys. J. C **76** (2016) 653, arXiv:1608.03953 [hep-ex].
- [45] D. J. Lange, Nucl. Instrum. Meth. A **462** (2001) 152.
- [46] R. D. Ball et al., Nucl. Phys. B **867** (2013) 244–289, arXiv:1207.1303 [hep-ph].
- [47] ATLAS Collaboration, Atl-phys-pub-2014-021, <https://cds.cern.ch/record/1966419>.
- [48] LHC Higgs Cross Section Working Group Collaboration, D. de Florian et al., arXiv:1610.07922 [hep-ph].
- [49] T. Gleisberg et al., JHEP **02** (2009) 007, arXiv:0811.4622 [hep-ph].
- [50] ATLAS Collaboration, Atl-phys-pub-2016-002, <http://cds.cern.ch/record/2119986>.
- [51] H.-L. Lai et al., Phys. Rev. D **82** (2010) 074024, arXiv:1007.2241 [hep-ph].
- [52] LHC Higgs Cross Section Working Group Collaboration, LHC Higgs Cross Section Working Group, arXiv:1201.3084 [hep-ph].
- [53] ATLAS Collaboration, JHEP **2014** (2014) 55, arXiv:1406.3660 [hep-ex].
- [54] M. Czakon and A. Mitov, Comput. Phys. Commun. **185** (2014) 2930, arXiv:1112.5675 [hep-ph].
- [55] P. Skands, Phys. Rev. D **82** (2010) 074018, arXiv:1005.3457 [hep-ph].
- [56] ATLAS Collaboration, G. Aad et al., Eur.Phys.J. **C70** (2010) 823–874, arXiv:1005.4568 [physics.ins-det].
- [57] A. Sherstnev and R. Thorne, Eur. Phys. J. C **55** (2008) 553–575, arXiv:0711.2473 [hep-ph].
- [58] ATLAS Collaboration, Atl-phys-pub-2016-005, <http://cds.cern.ch/record/2120826>.
- [59] G. Corcella et al., JHEP **01** (2001) 010, arXiv:hep-ph/0011363.
- [60] J. Pumplin et al., JHEP **07** (2002) 012, arXiv:hep-ph/0201195.
- [61] J. Alwall, M. Herquet, F. Maltoni, O. Mattelaer, and T. Stelzer, JHEP **06** (2011) 128, arXiv:1106.0522 [hep-ph].
- [62] LHC SUSY Cross Section Working Group, 2015, <https://twiki.cern.ch/twiki/bin/view/LHCPhysics/SUSYCrossSections>.
- [63] S. Carrazza, S. Forte, and J. Rojo, *Parton distributions and event generators*, pp. , 89–96. 2013. arXiv:1311.5887 [hep-ph].
- [64] L. Lönnblad and S. Prestel, JHEP **03** (2012) 019, arXiv:1109.4829 [hep-ph].
- [65] C. Borschensky et al., Eur. Phys. J. C **74** (2014) 3174, arXiv:1407.5066 [hep-ph].

- [66] ATLAS Collaboration, ATL-PHYS-PUB-2015-026, 2015,
<https://cds.cern.ch/record/2037717>.
- [67] M. Cacciari and G. P. Salam, Phys. Lett. B **659** (2008) 119–126,
[arXiv:0707.1378 \[hep-ph\]](https://arxiv.org/abs/0707.1378).
- [68] ATLAS Collaboration, ATLAS Collaboration, [arXiv:1703.09665 \[hep-ex\]](https://arxiv.org/abs/1703.09665).
- [69] ATLAS Collaboration, Atlas-conf-2014-018,
<https://cds.cern.ch/record/1700870>.
- [70] ATLAS Collaboration, ATLAS Collaboration, JINST **11** (2016) P04008,
[arXiv:1512.01094 \[hep-ex\]](https://arxiv.org/abs/1512.01094).
- [71] ATLAS Collaboration, ATL-PHYS-PUB-2015-022, 2015,
<https://cds.cern.ch/record/2037697>.
- [72] ATLAS Collaboration, ATL-PHYS-PUB-2016-012, 2016,
<https://cds.cern.ch/record/2160731>.
- [73] ATLAS Collaboration, ATLAS Collaboration, Eur. Phys. J. C **76** (2016) 292,
[arXiv:1603.05598 \[hep-ex\]](https://arxiv.org/abs/1603.05598).
- [74] ATLAS Collaboration, Eur. Phys. J. C **76** (2016) 666, [arXiv:1606.01813 \[hep-ex\]](https://arxiv.org/abs/1606.01813).
- [75] ATLAS Collaboration, ATL-PHYS-PUB-2015-027, 2015,
<https://cds.cern.ch/record/2037904>.
- [76] ATLAS Collaboration, ATL-PHYS-PUB-2015-023, 2015,
<https://cds.cern.ch/record/2037700>.
- [77] R. D. Cousins, K. E. Hymes, and J. Tucker, Nucl.Instrum.Meth.Phys.Res. **A612** (2010) 388–398, [arXiv:0905.3831 \[physics.data-an\]](https://arxiv.org/abs/0905.3831).
- [78] ATLAS Collaboration, ATLAS-CONF-2017-021, 2017,
<https://cds.cern.ch/record/2258143>.
- [79] ATLAS Collaboration, Atlas-conf-2016-037,
<http://cds.cern.ch/record/2205745>.
- [80] ATLAS Collaboration, Atlas-conf-2014-032,
<https://cds.cern.ch/record/1706245>.
- [81] G. Cowan, K. Cranmer, E. Gross, and O. Vitells, Eur. Phys. J. **C71** (2011) 1554, [arXiv:1007.1727 \[physics.data-an\]](https://arxiv.org/abs/1007.1727), [Erratum : Eur. Phys. J.C73,2501(2013)].
- [82] A. L. Read, Journal of Physics G : Nuclear and Particle Physics **28** (2002) 2693, <http://stacks.iop.org/0954-3899/28/i=10/a=313>.
- [83] F. James.
- [84] L. Moneta, K. Belasco, K. S. Cranmer, S. Kreiss, A. Lazzaro, D. Piparo, G. Schott, W. Verkerke, and M. Wolf, PoS **ACAT2010** (2010) 057, [arXiv:1009.1003 \[physics.data-an\]](https://arxiv.org/abs/1009.1003).
- [85] I. Antcheva et al., Comput. Phys. Commun. **180** (2009) 2499–2512, [arXiv:1508.07749 \[physics.data-an\]](https://arxiv.org/abs/1508.07749).

- [86] J. Neyman and E. S. Pearson, *Philosophical Transactions of the Royal Society of London A: Mathematical, Physical and Engineering Sciences* **231** (1933) 289–337,
<http://rsta.royalsocietypublishing.org/content/231/694-706/289.full.pdf>,
<http://rsta.royalsocietypublishing.org/content/231/694-706/289>.
- [87] S. S. Wilks, *Annals Math. Statist.* **9** (1938) 60–62.
- [88] A. Wald, *Transactions of the American Mathematical Society* **54** (1943) 426–482, <http://www.jstor.org/stable/1990256>.
- [89] I. Asimov, *Franchise*. Creative Education, 1 ed., May, 1989.
- [90] M. Baak et al., *Eur. Phys. J. C* **75** (2015) 153, [arXiv:1410.1280 \[hep-ex\]](https://arxiv.org/abs/1410.1280).
- [91] K. Cranmer, G. Lewis, L. Moneta, A. Shibata, and W. Verkerke, Cern-open-2012-016, <http://cds.cern.ch/record/1456844>.
- [92] W. Verkerke and D. P. Kirkby, eConf **C0303241** (2003) MOLT007, [arXiv:physics/0306116 \[physics\]](https://arxiv.org/abs/physics/0306116), [,186(2003)].
- [93] A. L. Read, *J. Phys. G* **28** (2002) 2693–2704.
- [94] J. Ellis, T. Falk, K. A. Olive, and Y. Santoso, *Nucl. Phys. B* **652** (2003) 259–347, [arXiv:hep-ph/0210205](https://arxiv.org/abs/hep-ph/0210205).
- [95] J. Ellis, K. A. Olive, and Y. Santoso, *Phys. Lett. B* **539** (2002) 107–118, [arXiv:hep-ph/0204192](https://arxiv.org/abs/hep-ph/0204192).
- [96] ATLAS Collaboration, ATLAS Collaboration, *Eur. Phys. J. C* **76** (2016) 259, [arXiv:1602.09058 \[hep-ex\]](https://arxiv.org/abs/1602.09058).
- [97] ATLAS Collaboration, ATLAS Collaboration, *Phys. Lett. B* **757** (2016) 334–355, [arXiv:1602.06194 \[hep-ex\]](https://arxiv.org/abs/1602.06194).