

# Measurement of the Higgs boson transverse momentum spectrum in the $WW$ decay channel at 8 TeV and first results at 13 TeV

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# Abstract

The cross section for Higgs boson production in pp collisions is studied using the  $H \rightarrow W^+W^-$  decay mode, followed by leptonic decays of the W bosons, leading to an oppositely charged electron-muon pair in the final state. The measurements are performed using data collected by the CMS experiment at the LHC with pp collisions at a centre-of-mass energy of 8 TeV, corresponding to an integrated luminosity of  $19.4\text{fb}^{-1}$ . The Higgs boson transverse momentum ( $p_T$ ) is reconstructed using the lepton pair  $p_T$  and missing  $p_T$ . The differential cross section times branching fraction is measured as a function of the Higgs boson  $p_T$  in a fiducial phase space defined to match the experimental acceptance in terms of the lepton kinematics and event topology. The production cross section times branching fraction in the fiducial phase space is measured to be  $39 \pm 8$  (stat)  $\pm 9$  (syst) fb. The measurements are compared to theoretical calculations based on the standard model to which they agree within experimental uncertainties.



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## Chapter 1.

# Electroweak and QCD physics at LHC



## Chapter 2.

# The CMS experiment at the LHC

### 2.1. The Large Hadron Collider

### 2.2. The CMS experiment

### 2.3. The CMS trigger system

### 2.4. Objects definition and event reconstruction



## Chapter 3.

# Higgs boson properties in the $H \rightarrow WW$ decay channel

### 3.1. Higgs boson measurements at LHC

The discovery of a new boson consistent with the standard model (SM) Higgs boson has been reported by ATLAS and CMS Collaborations in 2012. The discovery has been followed by a comprehensive set of studies of properties of this new boson in several production and decay channels and no evidence of deviation from the SM expectation has been found so far. The CMS studies in the  $H \rightarrow WW \rightarrow 2\ell 2\nu$  decay channel include the measurement of the Higgs properties, as well as constraints on the Higgs total decay width and gauge bosons anomalous couplings.

### 3.2. Higgs boson measurements in the $H \rightarrow WW$ decay channel



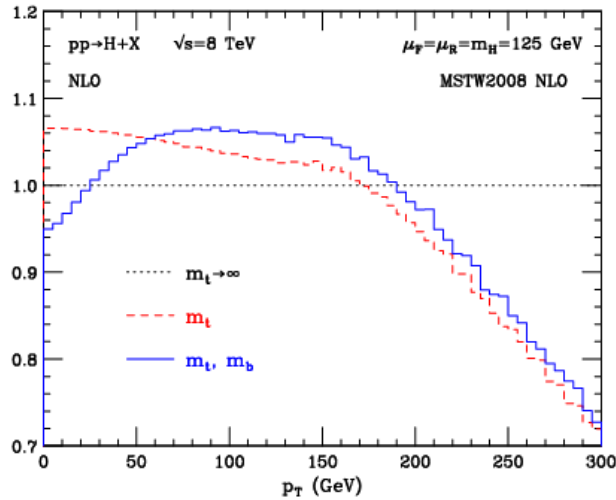
## Chapter 4.

# Measurement of the Higgs boson transverse momentum at 8 TeV using $H \rightarrow WW \rightarrow 2\ell 2\nu$ decays

### 4.1. Introduction

The Higgs boson production at hadron colliders is characterized by  $p_T^H$  and  $\eta$ . The  $\eta$  distribution is essentially driven by the PDF of the partons in the colliding hadrons, and it is only mildly sensitive to radiative corrections. The  $p_T^H$  distribution is instead sensitive to QCD radiative corrections. Considering the ggH production mode, at LO in perturbation theory,  $\mathcal{O}\alpha_s^2$ , the Higgs boson is always produced with  $p_T^H$  equal to zero. Indeed in order to have  $p_T$  different from zero, the Higgs boson has to recoil at least against one parton. Higher order corrections to the ggH process are numerically large and are known at NLO including full top quark mass dependence [1, 2], and at NNLO using the so-called large- $m_t$  approximation [3, 4, 5], in which the top quark mass is assumed to be very large and the fermionic loop is replaced by an effective vertex of interaction. Starting from the NLO, the Higgs boson can be produced recoiling against other final state partons, resulting in a finite  $p_T^H$ . For this reason the LO process for Higgs production at  $p_T \neq 0$  is at  $\mathcal{O}\alpha_s^3$ , and the counting of perturbative orders differs between inclusive Higgs boson production and  $p_T^H$  distribution. Also, NNLO QCD corrections in the  $p_T^H$  observable have recently been shown [6].

When  $p_T^H \sim m_H$  the QCD radiative corrections to  $p_T^H$  differential cross section are theoretically evaluated using fixed-order calculations. When  $p_T^H \ll m_H$  the perturbative expansion does not converge due to the presence of large logarithmic terms of the form  $\alpha_s^n \ln^{2n} m_H^2/p_T^2$ , leading to a divergence of  $d\sigma/dp_T$  in the limit of  $p_T \rightarrow 0$ . For computing the  $p_T^H$  spectrum in this region soft-gluon resummation techniques are used, and matched to the fixed-order calculation in the  $p_T^H \sim m_H$  region. For the  $p_T^H$  differential cross section the large- $m_t$  calculation is a crude approximation, since it is known that the top quark mass has a non-negligible effect on the shape of the spectrum. Moreover the inclusion of the bottom quark contribution in the fermionic loop can significantly modify the  $p_T^H$  shape [7], as shown in Fig. 4.1. Hence, a precise experimental measurement of the  $p_T^H$  spectrum is important to test the existing SM calculations.



**Figure 4.1.:**  $p_T^H$  distribution computed at NLO ( $\alpha_s^3$ ) and normalized to the calculation obtained in the large- $m_t$  approximation. The red dashed line corresponds to the calculation including the top quark mass while the blue line refers to the calculation including also the bottom quark effects.

Possible extensions of the SM predict a modification of the Higgs boson couplings to gluons and to the top quark. Many of these models actually predict the existence of new states that interact with the SM Higgs boson, but are beyond the direct production reach at the actual LHC energies. The effect of these new states could however show up as a deviation of the Higgs boson couplings with respect to the SM expectation. The modification of the couplings, as shown in Refs. [8, 9], can change the kinematics of the Higgs boson production and the effect can be particularly sizeable in the tail of the  $p_T^H$  distribution. Other models, such as Composite Higgs [10], predict the existence of top-partners, which are heavy resonances with the same quantum numbers as the top quark, that can interact



with the Higgs boson in the ggH fermionic loop, changing the  $p_T^H$  shape with respect to what the SM predicts [11]. The measurement of the  $p_T^H$  spectrum is thus a useful tool for indirect searches of new particles predicted by theories beyond the SM.

Measurements of the fiducial cross sections and of several differential distributions, using the  $\sqrt{s} = 8$  TeV LHC data, have been reported by ATLAS [12, 13, 14] and CMS [15, 16] for the  $H \rightarrow ZZ \rightarrow 4\ell$  ( $\ell = e, \mu$ ) and  $H \rightarrow \gamma\gamma$  decay channels. In this chapter a measurement of the fiducial cross section times branching fraction ( $\sigma \times \mathcal{B}$ ) and  $p_T$  spectrum for Higgs boson production in  $H \rightarrow WW \rightarrow e^\pm \mu^\mp \nu \nu$  decays, based on  $\sqrt{s} = 8$  TeV LHC data, is reported.

The analysis is performed looking at different flavour leptons in the final state in order to suppress the sizeable contribution of backgrounds containing a same-flavour lepton pair originating from Z boson decay.

Although the  $H \rightarrow WW \rightarrow 2\ell 2\nu$  channel has lower resolution in the  $p_T^H$  measurement compared to the  $H \rightarrow \gamma\gamma$  and  $H \rightarrow ZZ \rightarrow 4\ell$  channels because of neutrinos in the final state, the channel has a significantly larger  $\sigma \times B$ , exceeding those for  $H \rightarrow \gamma\gamma$  by a factor of 10 and  $H \rightarrow ZZ \rightarrow 4\ell$  by a factor of 85 for a Higgs boson mass of 125 GeV [17], and is characterized by good signal sensitivity. Such sensitivity allowed the observation of a Higgs boson at the level of 4.3 (5.8 expected) standard deviations for a mass hypothesis of 125.6 GeV using the full LHC data set at 7 and 8 TeV [18].

The measurement is performed in a fiducial phase space defined by kinematic requirements on the leptons that closely match the experimental event selection.

The effect of the limited detector resolution, as well as the selection efficiency with respect to the fiducial phase space are corrected to particle level with an unfolding procedure [19], as explained in Sec. 4.7.

## 4.2. Datasets, Triggers and MC samples

This analysis relies on the published  $H \rightarrow WW$  measurements [18] in terms of code, selections and background estimates for both the gluon fusion (ggH) [20] and the vector boson fusion (VBF) [AN-13-097] production mechanisms.

### 4.2.1. Datasets and triggers

The datasets used for the analysis correspond to  $19.4\text{fb}^{-1}$  at  $\sqrt{s} = 8$  TeV of integrated luminosity composed of the following CMS data taking periods during 2012: 2012A ( $892\text{ pb}^{-1}$ ), 2012B ( $4440\text{ pb}^{-1}$ ), and 2012C ( $6898\text{ pb}^{-1}$ ) and 2012D ( $7238\text{ pb}^{-1}$ ). Data have been checked and validated and only data corresponding to good data taking quality are considered. The  $e^\pm\mu^\mp$  final state is considered in this analysis.

For the data samples, the events are required to fire one of the unprescaled single-electron, single-muon or muon-electron triggers. A full description of these triggers is given in [21] for 8 TeV data. Although identification and isolation criteria are also applied, a brief overview of the HLT transverse momentum ( $p_T$ ) criteria on the leptons is given in Table 4.1. While the HLT lepton  $p_T$  thresholds of 17 and 8 GeV for the double lepton triggers accommodate the offline lepton  $p_T$  selection of 20 and 10 GeV, the higher  $p_T$  thresholds in the single lepton triggers help partially recovering double lepton trigger inefficiencies as a high  $p_T$  lepton is on average expected due to the kinematic of the Higgs decay.

**Table 4.1.:** Highest transverse momentum thresholds applied in the lepton triggers at the HLT level. Double set of thresholds indicates the thresholds for each leg of the double lepton triggers.

Trigger Path	7 TeV	8 TeV
Single-Electron	$p_T > 27\text{ GeV}$	$p_T > 27\text{ GeV}$
Single-Muon	$p_T > 15\text{ GeV}$	$p_T > 24\text{ GeV}$
Muon-Electron	$p_T > 17\text{ and } 8\text{ GeV}$	$p_T > 17\text{ and } 8\text{ GeV}$
Electron-Muon	$p_T > 17\text{ and } 8\text{ GeV}$	$p_T > 17\text{ and } 8\text{ GeV}$

No trigger requirement is made on the simulated events but the combined trigger efficiency is estimated from data and applied as a weight to all simulated events. The detailed trigger efficiencies and the weighting procedure can be found in Appendix C of [20] [22]. The average trigger efficiency for signal events that pass the full event selection is measured to be about 96% in the  $e\mu$  final state for a Higgs boson mass of about 125 GeV.

### 4.2.2. Monte-Carlo samples

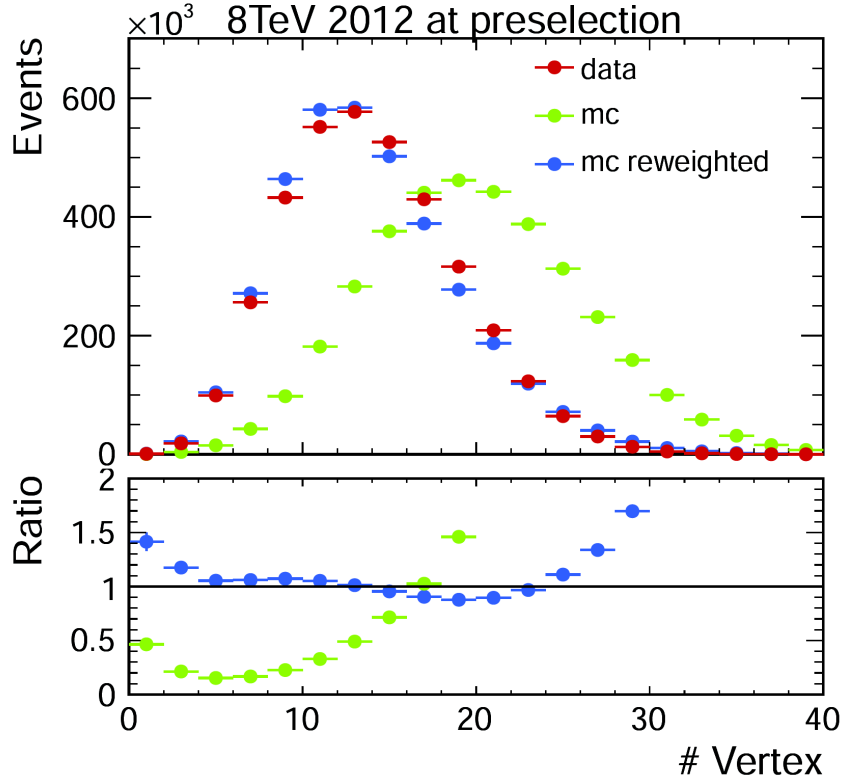
Several Monte Carlo event generators are used to simulate the signal and background processes:

- The POWHEG program [23] provides event samples for the  $H \rightarrow WW$  signal for the gluon fusion ( $ggH$ ) and VBF production mechanisms, as well as  $t\bar{t}$  and  $tW$  processes.
- The  $q\bar{q} \rightarrow WW$ , Drell-Yan,  $ZZ$ ,  $WZ$ ,  $W\gamma$ ,  $W\gamma^*$ , tri-bosons and  $W$ +jets processes are generated using the MADGRAPH 5.1.3 [24] event generator.
- The  $VH$  process is simulated using PYTHIA 6.424 [25].

For leading-order generators samples, the CTEQ6L [26] set of parton distribution functions (PDF) is used, while CT10 [27] is used for next-to-leading order (NLO) ones. Cross section calculations [**LHCHiggsCrossSectionWorkingGroup:2011ti**] at next-to-next-to-leading order (NNLO) are used for the  $H \rightarrow WW$  process (POWHEG NLO generator is tuned to reproduce NNLO accuracy on the on-shell Higgs  $p_T$  spectrum and scaled to NNLO inclusive cross-section), while NLO calculations are used for background cross sections. For all processes, the detector response is simulated using a detailed description of the CMS detector, based on the GEANT4 package [28].

Minimum bias events are superimposed on the simulated events to emulate the additional  $pp$  interactions per bunch crossing (pile-up). The number of pile-up events simulated in the MC samples (in the same bunch crossing, in time, or in the previous or following one, out of time pile-up) have been generated poissonianly sampling from a distribution similar to what is expected from data. These samples are reweighted to represent the pile-up distribution as measured in the data. For a given range of analyzed runs, the mean number of pile-up interactions per bunch crossing is estimated per luminosity block using the instantaneous luminosity provided by the LHC, integrated over the entire run range and normalized. This distribution is then used to reweight the simulated pile-up distribution. The average number of pile-up events per beam crossing in the 2011 data is about 10, and in the 2012 data it is about 20.

The contribution of the  $t\bar{t}H$  production mechanisms was checked to be negligible in each bin of  $p_T^H$  (below 1%) and was not taken into account. In figure 4.3 is shown the relative fraction of the four different production modes in each bin of  $p_T^H$ .



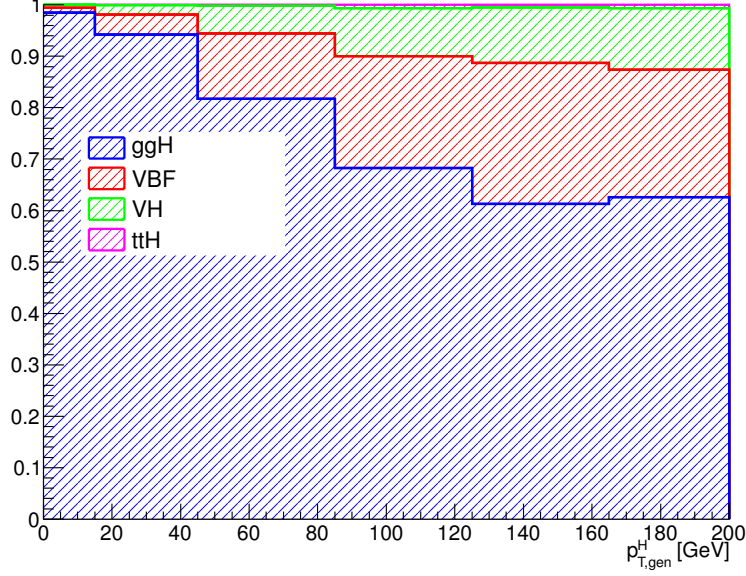
**Figure 4.2.:** Distribution of the number of vertices in data and in simulation, before and after applying the pile-up reweighting.

### 4.3. Analysis Strategy

The analysis presented here is based on that used in the previously published  $H \rightarrow WW \rightarrow 2\ell 2\nu$  measurements by CMS [18], modified to be inclusive in the number of jets. This modification significantly reduces the uncertainties related to the modelling of the number of jets produced in association with the Higgs boson.

#### 4.3.1. Event reconstruction and selections

The electron selection is based on two multivariate discriminants, one specialised in identifying the electron object and the other for isolation. The cut value for each discriminant is optimised to provide a good fake electron rejection and to improve the signal acceptance.



**Figure 4.3.:** Relative fraction of ggH, VBF, VH and ttH in each bin of the Higgs boson transverse momentum.

Muons are reconstructed using the standard CMS selection and are required to be identified both in the tracker (*tracker muon*) and in the muon chambers (*global muon*). Additionally quality criteria on the muon track are required, such as to have at least 10 hits in the tracker (at least one of which in the pixel detector) and to have  $\chi^2/ndf < 10$ . Muon isolation is based on the Particle-Flow algorithm. An MVA approach is considered, based on the radial distributions of the Particle-Flow candidates inside a cone of radius 0.5 around the muon direction.

The efficiencies for the identification and isolation of the electrons and muons are measured in data and in simulation selecting a pure sample of leptons coming from the  $Z \rightarrow \ell\ell$  decay. The measured efficiencies are used as scale factors to correct the MC simulation to precisely model the data. Similarly, the trigger efficiency extracted from data is applied to MC samples to correct for the additional loss.

Jets in this analysis are reconstructed by combining the energy measured in the calorimeters and tracks from charged particles on basis of the standard CMS particle flow algorithm and using the anti- $k_T$  clustering algorithm with  $R = 0.5$ . Events will be classified into zero jet, one jet and VBF topologies by counting jets within  $|\eta| < 4.7$  and for  $p_T > 30$  GeV.

Here I could add some details about jets (see AN/2012-194) if they are not already discussed in the objects section.

Background events from  $t\bar{t}$  and single-top production are rejected applying a soft-muon veto and b-tagging veto. The former selection requires that in the event there are no muons from b-decays passing the following cuts:

- the muon is reconstructed as TrackerMuon (and passes the TMLastStationTight ID);
- the number of hits of the muon in the Silicon Tracker is greater than 10;
- the transverse impact parameter of the muon is less than 0.2 cm;
- if  $p_T > 20$  GeV then the muon is required to be non-isolated with  $ISO/p_T > 0.1$ .

The latter veto rejects events that contain jets tagged as b-jets using two different algorithms for high and low  $p_T$  jets. For jets with  $p_T$  between 10 and 30 GeV, the Track-Counting-High-Efficiency (TCHE) algorithm, with a cut at 2.1 on the discriminating variable, is applied. For jets above 30 GeV, a more performant algorithm, Jet-Probability (JP), is used. Jets are identified as b-jets by the JP algorithm if the discriminating variable has a value above 1.4. In the following a b-tagged jet is defined as a jet, within  $|\eta| < 2.4$  (b-tagging requires the tracker information), with a value of the discriminating variable above the mentioned thresholds for the two algorithms.

The event selection consists of several steps. The first step is to select WW-like events applying a selection that is heavily based on the main analysis selection except for few different cuts explained below. The WW-like event preselection consists of the following set of cuts:

**1. Lepton preselection:**

- at least two opposite-sign and opposite-flavour ( $e\mu$ ) leptons reconstructed in the event;
- $|\eta| < 2.5$  for electrons and  $|\eta| < 2.4$  for muons;
- $p_T > 20$  GeV for the leading lepton. For the trailing lepton, the transverse momentum is required to be larger than 10 GeV.

**2. Extra lepton veto:** the event is required to have two and only two opposite-sign leptons passing the lepton selection.

- 
3.  **$E_T^{\text{miss}}$  preselection:** particle flow  $E_T^{\text{miss}}$  is required to be greater than 20 GeV.
  4. **Di-lepton mass cut:**  $m_{\ell\ell} > 12$  GeV in order to reject low mass resonances and QCD backgrounds.
  5. **Di-lepton  $p_T$  cut:**  $p_T^{\ell\ell} > 30$  GeV.
  6. **projected  $E_T^{\text{miss}}$  selection:** minimum projected  $E_T^{\text{miss}}$  required to be larger than 20 GeV.
  7. **Transverse mass:**  $m_T > 60$  GeV to reject Drell-Yan to  $\tau\tau$  events.

In addition to the WW-like preselection other cuts are applied in order to reduce the top background ( $t\bar{t}$  and single-top), which is one of the main backgrounds in this final state. We operate two different selections depending on the number of jets with  $p_T > 30$  GeV in the event. This is done to suppress the top background both in the low  $p_T^H$  region, where 0-jets events have the biggest contribution, and for higher values where also larger jet multiplicity events are important. The selection for 0-jets events relies on a soft muon veto, which rejects events with non-isolated soft muons (likely belonging to b-jets), and on a soft jets (with  $p_T < 30$  GeV) anti b-tagging requirement. The latter requirement exploits the Track Counting High Efficiency tagger (TCHE) to reject soft jets that are likely to come from b quarks hadronization.

For events with a jet multiplicity greater or equal than one, a different selection is applied. In this case we exploit the good b-tagging performances of the *JetBProbability* tagger to reject all the jets with  $p_T > 30$  GeV that are likely to come from a b quark. This jet veto relies on a cut on the *JetBProbability* tagger discriminant. Any jet with a discriminant value below 1.4 is identified as a non b-jet. The analysis selection requires to have no events containing b-tagged jets with  $p_T > 30$  GeV.

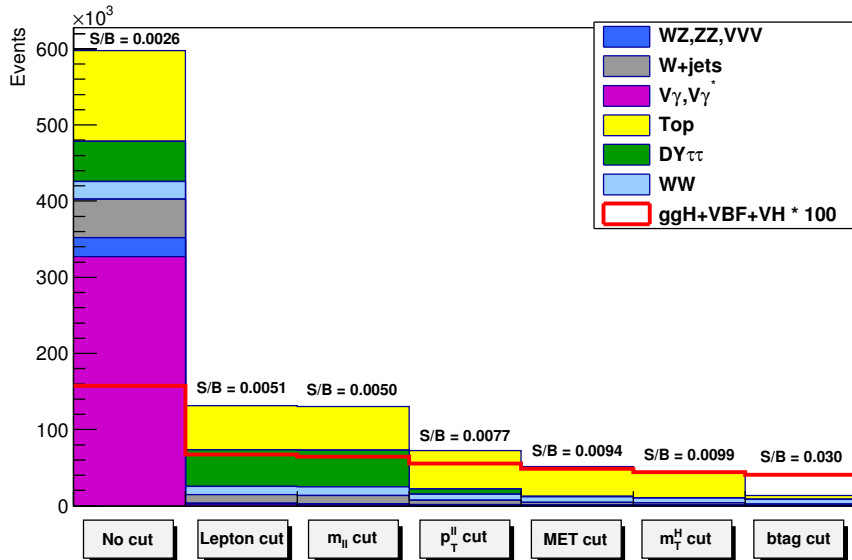
A cut-flow plot is reported in figure 4.4 showing the effect of each selection on top of Monte Carlo samples. In the first bin, labelled as *No cut*, no selection has been applied and the bin content correspond to the total expected number of events with a luminosity of  $19.46 \text{ fb}^{-1}$ . All the events in this bin have at least two leptons with a loose transverse momentum cut of 8 GeV. In the following bin the lepton cuts are applied, including the requirement to have two opposite-sign and opposite-flavour leptons and the extra lepton veto. Then are progressively reported all the other selections, showing the effect of each

cut on backgrounds and signal. For each selection is also reported the expected signal over background ratio which after the full selection reach a maximum value around 3%.

### 4.3.2. Fiducial phase space

The Higgs boson transverse momentum is measured in a fiducial phase space, whose requirements are chosen in order to minimize the dependence of the measurements on the underlying model of the Higgs boson properties and its production mechanism.

The exact requirements are determined by considering the two following correlated quantities: the reconstruction efficiency for signal events originating from within the fiducial phase space (fiducial signal efficiency  $\epsilon_{\text{fid}}$ ), and the ratio of the number of reconstructed signal events that are from outside the fiducial phase space (“out-of-fiducial” signal events) to the number from within the fiducial phase space. The requirement of having a small fraction of out-of-fiducial signal events, while at the same time preserving a high value of the fiducial signal efficiency  $\epsilon_{\text{fid}}$ , leads to a loosening of the requirements on the low-resolution variables,  $E_{\text{T}}^{\text{miss}}$  and  $m_{\text{T}}$ , with respect to the analysis selection.



**Figure 4.4.:** Effect of single selections on MC samples. The signal (red line) is multiplied by 100 and superimposed on stacked backgrounds. In each bin, corresponding to a different selection, is reported the expected number of events in MC at a luminosity of  $19.46 \text{ fb}^{-1}$ .



The fiducial phase space used for the cross section measurements is defined at the particle level by the requirements given in Table 4.2. The leptons are defined as Born-level leptons, i.e. before the emission of final-state radiation (FSR), and are required not to originate from leptonic  $\tau$  decays. The effect of including FSR is evaluated to be of the order of 5% in each  $p_T^H$  bin. For the VH signal process the two leptons are required to originate from the  $H \rightarrow WW \rightarrow 2\ell 2\nu$  decays in order to avoid including leptons coming from the associated W or Z boson.

**Table 4.2.:** Summary of requirements used in the definition of the fiducial phase space.

Physics quantity	Requirement
Leading lepton $p_T$	$p_T > 20 \text{ GeV}$
Subleading lepton $p_T$	$p_T > 10 \text{ GeV}$
Pseudorapidity of electrons and muons	$ \eta  < 2.5$
Invariant mass of the two charged leptons	$m_{\ell\ell} > 12 \text{ GeV}$
Charged lepton pair $p_T$	$p_T^{\ell\ell} > 30 \text{ GeV}$
Invariant mass of the leptonic system in the transverse plane	$m_T^{\ell\ell\nu} > 50 \text{ GeV}$
$E_T^{\text{miss}}$	$E_T^{\text{miss}} > 0$

A detailed description of the fiducial region definition and its optimization is given in appendix A.

### 4.3.3. Binning of the $p_T^H$ distribution

Experimentally, the Higgs boson transverse momentum is reconstructed as the vector sum of the lepton momenta in the transverse plane and  $E_T^{\text{miss}}$ .

$$\vec{p}_T^H = \vec{p}_T^{\ell\ell} + \vec{p}_T^{\text{miss}} \quad (4.1)$$

Compared to other differential analysis of the Higgs cross section, such as those in the ZZ and  $\gamma\gamma$  decay channels, this analysis has to cope with the limited resolution due to the  $E_T^{\text{miss}}$  entering the transverse momentum measurement. The effect of the limited  $E_T^{\text{miss}}$  resolution has two main implications on the analysis strategy. The first one is that the choice of the binning in the  $p_T^H$  spectrum needs to take into account the detector resolution. The

second implication is that migrations of events across bins are significant and an unfolding procedure needs to be applied to correct for selection efficiencies and bin migration effects.

Given these aspects the criterion that was used to define the  $p_T^H$  bin size is devised to keep under control the bin migrations due to the finite resolution.

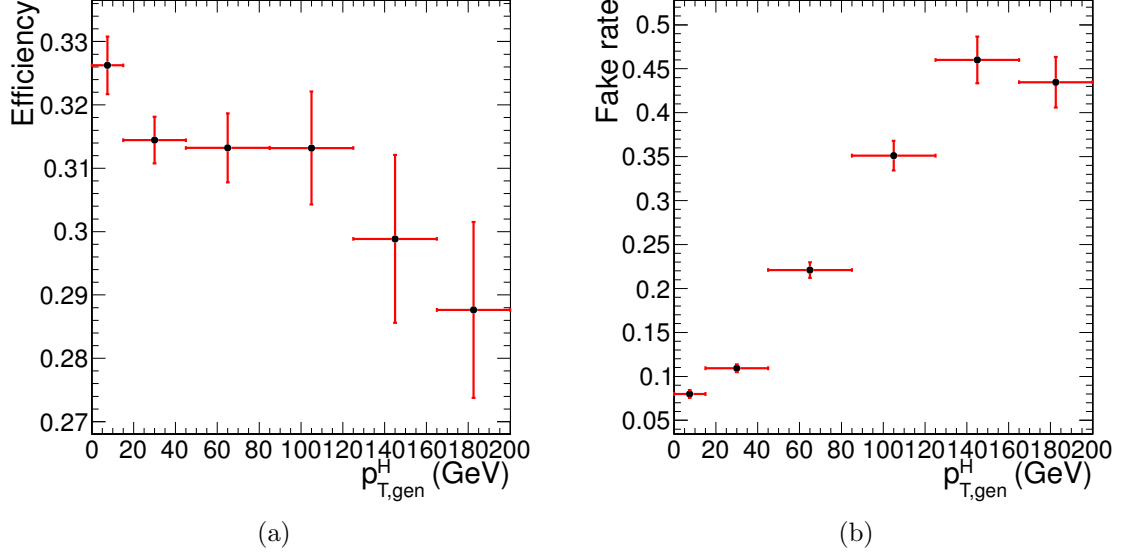
For any given bin  $i$  we can define the purity  $P_i$  on a signal sample as the number events that are generated and also reconstructed in that bin,  $N_i^{\text{GEN|RECO}}$ , divided by the number of events reconstructed there  $N_i^{\text{RECO}}$ :

$$P_i = \frac{N_i^{\text{GEN|RECO}}}{N_i^{\text{RECO}}} \quad . \quad (4.2)$$

Where  $N_i^{\text{GEN|RECO}}$  is the number of events that are both generated and reconstructed in a  $p_T^H$  bin  $i$ , while  $N_i^{\text{RECO}}$  is the number of events that are reconstructed in bin  $i$ . We have chosen the bin width in such a way as to make the smallest bins able to ensure a purity of about 60% on a gluon fusion sample. Following this prescription we have divided the whole  $p_T^H$  range in six different bins: [0-15 GeV], [15-45 GeV], [45-85 GeV], [85-125 GeV], [125-165 GeV], [165- $\infty$  GeV].

The efficiency of the analysis selection with respect to the fiducial phase space is reported in Fig. 4.5 (a) for each  $p_T^H$  bin. The efficiency denominator is the number of events that are inside the fiducial phase space, while the numerator is the number of events that pass both the analysis and the fiducial phase space selections, in each  $p_T^H$  bin. The fake rate, defined by the ratio of signal events that pass the analysis selection but are not within the fiducial phase space, divided by the total number of events passing both the analysis and the fiducial phase space selections is shown in Fig. 4.5 (b). For both the selection efficiency and the fake rate, all the signal production mechanisms are included. The overall efficiency and fake rate are:  $\epsilon = 0.362 \pm 0.005$  and  $fake\ rate = 0.126 \pm 0.004$ , where the errors are only statistical.

If a  $4\pi$  acceptance is defined, requiring just that the Higgs decays to WW and then to  $2\ell 2\nu$ , the efficiency becomes  $\epsilon = 0.03960 \pm 0.00033$  and the fake rate is zero.



**Figure 4.5.:** Efficiency of the full selection (a) and fake rate (b) as a function of  $p_T^H$ .

## 4.4. Background estimation

Add plots for each background process

### 4.4.1. Top quark background

In this analysis the top quark background is divided into two different categories depending on the number of jets in the event. In the two categories different selections are applied, especially concerning the b-tagging requirements.

The general strategy for determining the residual top events in the signal region is to first measure the top tagging efficiencies from an orthogonal region of phase space in data. The orthogonal phase space is defined inverting the b-veto requirement of the signal region, in such a way to have a control region enriched in top quark events. Then, using this efficiency, the number of events with the associated uncertainty is propagated from the control region to the signal region. The number of surviving top events in the signal region would then be:

$$N_{bveto}^{signal} = N_{btag}^{control} \cdot \frac{1 - \epsilon_{top}}{\epsilon_{top}} \quad (4.3)$$

where  $N_{btag}^{control}$  is the number of events in the control region and  $\epsilon_{top}$  is the efficiency as measured in data.

The methods to estimate the top background contribution in the two jet categories are different and are explained below.

### 0-jets category

Most of the top background, composed of  $t\bar{t}$  and  $tW$  processes, is rejected in the 0-jet bin by the jet veto. The top-tagging efficiency in the zero jet bin,  $\epsilon_{tag}^{0-jet}$ , is the probability for a top event to fail one of either the b-tagging veto or the soft muon veto, and is defined as:

$$\epsilon_{tag} = \frac{N_{tag}^{control}}{N^{control}} \quad , \quad (4.4)$$

where  $N^{control}$  is the number of events in the top control phase space defined requiring one b-tagged jet with  $p_T > 30$  GeV, and  $N_{tag}^{control}$  is the subset of those events that pass either the soft muon tagging or the low- $p_T$  b jet tagging. The purity of this control sample, as estimated from simulation, is about 97%. The remaining 3% background contribution is estimated from simulation and subtracted from the numerator and denominator of Eq. (4.5). The efficiency  $\epsilon_{top}^{0-jet}$  can then be estimated using the following formula:

$$\epsilon_{top}^{0-jet} = f_{t\bar{t}} \cdot \epsilon_{2b} + f_{tW} \cdot (x \cdot \epsilon_{2b} + (1 - x) \cdot \epsilon_{tag}) \quad , \quad (4.5)$$

$$\epsilon_{2b} = 1 - (1 - \epsilon_{tag})^2 \quad , \quad (4.6)$$

where  $f_{t\bar{t}}$  and  $f_{tW}$  are the  $t\bar{t}$  and  $tW$  fractions respectively,  $x$  is the fraction of  $tW$  events containing 2 b jets, and  $\epsilon_{2b}$  is the efficiency for a top event with 0 counted jets, i.e. two soft b jets, to pass the top veto. For the ratio of  $t\bar{t}$  and  $tW$  cross-sections an uncertainty of 17% is assumed. The fraction  $f_{t\bar{t}}$  is estimated using MC simulation of the  $t\bar{t}$  and  $tW$  processes at NLO accuracy.

Using this procedure a data/simulation scale factor of  $0.98 \pm 0.17$  is found, and is applied to correct the MC simulation in order to match the data.

### Category with more than 0 jets

The strategy for the estimation of the top background in events with at least one jet with  $p_T$  greater than 30 GeV is the following. First of all the efficiency for tagging a b jet is measured both in data and simulation and the values are used to correct the simulation for different b-tagging efficiencies in data and simulation. This evaluation is performed in a control region, called CtrlTP, containing at least two jets, using a Tag&Probe technique. The procedure to extract these scale factors is presented in Sec. 4.4.1. Then a larger statistics control region, CtrlDD, is defined by requiring at least one b-tagged jet and we use the simulation, corrected for the previously computed b-tagging efficiency scale factor, to derive the factor that connects the number of events in CtrlDD to the number of events in the signal region. This second step is explained in detail in Sec. 4.4.1.

### Tag&Probe

The Tag&Probe technique is a method to estimate the efficiency of a selection on data. It can be applied whenever one has two objects in one event, by using one of the two, the *tag*, to identify the process of interest, and using the second, the *probe*, to actually measure the efficiency of the selection being studied. In our case we want to measure the b-tagging efficiency, so what we need is a sample with two b-jets per event. The easiest way to construct such a sample is to select  $t\bar{t}$  events.

The CtrlTP control region is defined selecting the events which pass the lepton preselection cuts listed in Sec. ??, and have at least two jets with  $p_T$  greater than 30 GeV. One of the two leading jets is required to have a *JetBProbability* score higher than 0.5. From events in this control region we built *tag-probe* pairs as follows. For each event the two

leading jets are considered. If the leading jet passes the *JetBProbability* cut of 0.5, that is considered a *tag*, and the sub-leading jet is the *probe*. In order to avoid any bias that could arise from the probe being always the second jet, the pair is tested also in reverse order, meaning that the sub-leading jet is tested against the *tag* selection, and in case it passes, then the leading jet is used as *probe* in an independent *tag-probe* pair. This means that from each event passing the CrtITP cuts one can build up to two *tag-probe* pairs.

If the *tag* selection were sufficient to suppress any non top events, one could estimate the efficiency by dividing the number of *tag-probe* pairs in which the *probe* passes the analysis cut *JetBProbability*  $> 1.4$  (*tag-pass-probe*) by the total number of *tag-probe* pairs. However this is not the case. In order to estimate the efficiency in the presence of background a variable that discriminates between true b-jets and other jets in a  $t\bar{t}$  sample is chosen. The variable is the  $p_T$  of the *probe* jet. For real b-jets this variable has a peak around 60 GeV, while it does not peak for other jets. The idea is to fit simultaneously the  $p_T$  spectrum for *probe* jets in *tag-pass-probe* and *tag-fail-probe* pairs, linking together the normalizations of the two samples as follows:

$$N_{TPP} = N_s \epsilon_s + N_b \epsilon_b \quad (4.7)$$

$$N_{TFP} = N_s(1 - \epsilon_s) + N_b(1 - \epsilon_b) \quad (4.8)$$

where  $N_{TPP}$  is the number of *tag-pass-probe* pairs,  $N_{TFP}$  is the number of *tag-fail-probe* pairs,  $N_s$  is the number of *tag-probe* pairs in which the probe is a b-jet,  $N_b$  is the number of *tag-probe* pairs in which the probe is a not b-jet,  $\epsilon_s$  is the b-tagging efficiency,  $\epsilon_b$  is the probability of identifying as b-jet a non-b-jets, i.e. the mistag rate.

A  $\chi^2$  simultaneous fit of the *probe*  $p_T$  spectrum for *tag-pass-probe* and *tag-fail-probe* pairs is performed, deriving the shapes for true b-jets and non-b-jets from the simulation, and extracting  $N_s$ ,  $N_b$ ,  $\epsilon_s$  and  $\epsilon_b$  from the fit. The result of the fit on simulation is shown in Fig. 4.6. The relevant efficiencies are:

$$\epsilon_s^{MC} = 0.7663 \pm 0.0072 \quad (4.9)$$

$$\epsilon_b^{MC} = 0.208 \pm 0.015 \quad (4.10)$$

We have checked that these values are consistent with the true value for the b-tagging efficiency. The true value is computed by selecting jets that are matched within a cone of  $\Delta R < 0.5$  with a generator level b-quark, and counting the fraction of those that pass the *JetBProbability* cut of 1.4. This means that the *tag-probe* method does not introduce biases within the simulation statistic accuracy.

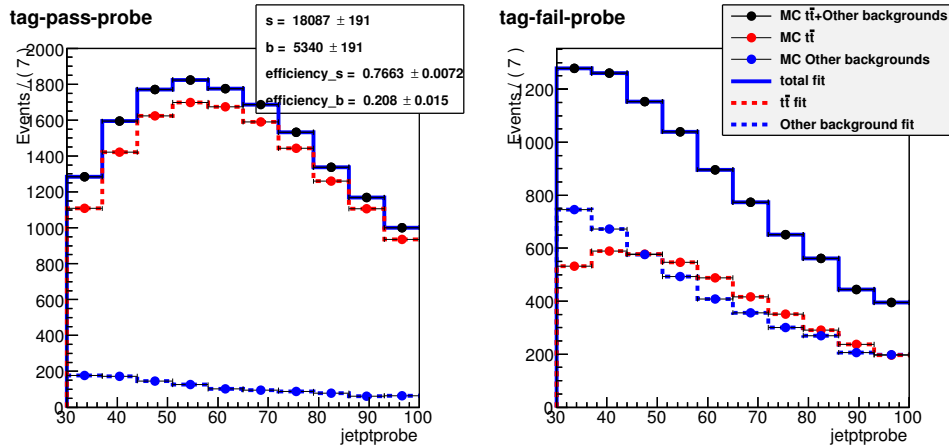
In order to assess the robustness of the fit, 5000 toy MC samples have been generated with a statistics equivalent to the one expected in data and the same fit is performed. All the 5000 fit succeeded, and the pull distributions for  $\epsilon_s$  and  $\epsilon_b$  parameters are shown in Fig. 4.7. The plots show the pull of the efficiencies measured in the fit, where the pull variable for each toy  $i$  is defined as:

$$\text{pull}(\epsilon_{s(b)}) = \frac{\epsilon_{s(b)}^{\text{true}} - \epsilon_{s(b)}^i}{\sigma(\epsilon_{s(b)}^i)} \quad (4.11)$$

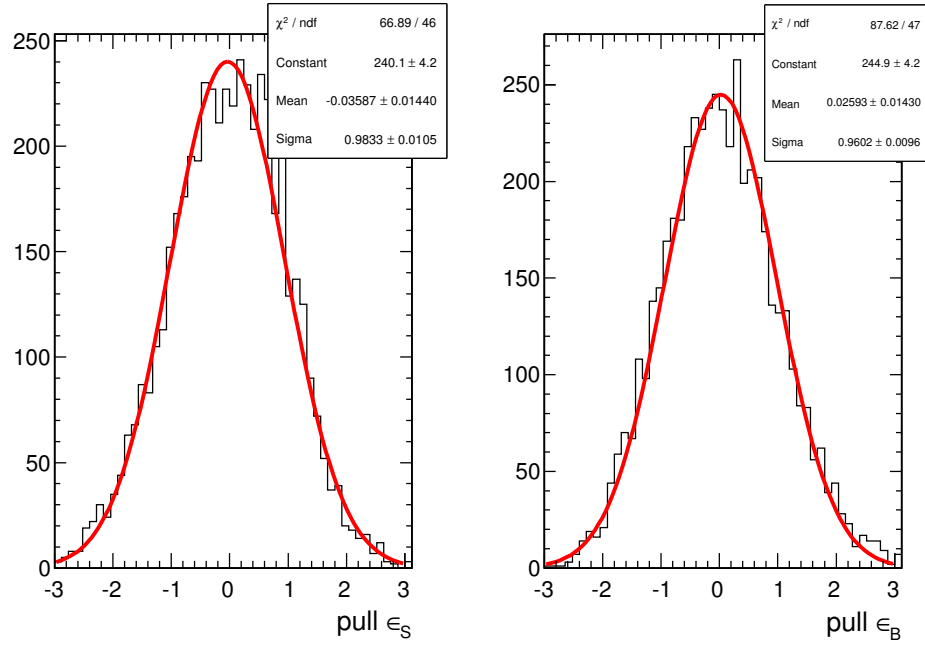
The pulls are centered on 0 and have  $\sigma$  close to 1, as expected.

An example fit for one of the toys is shown in Fig. 4.8

Before running the fit on data, the shapes used in the fit have been validated. To do so, a purer top enriched phase space has been defined by requiring exactly two jets with *JetBProbability* score higher than 1.5 and no additional b-tagged jets, rejecting also jets with  $p_T$  smaller than 30 GeV. On this purer sample we have compared data against the

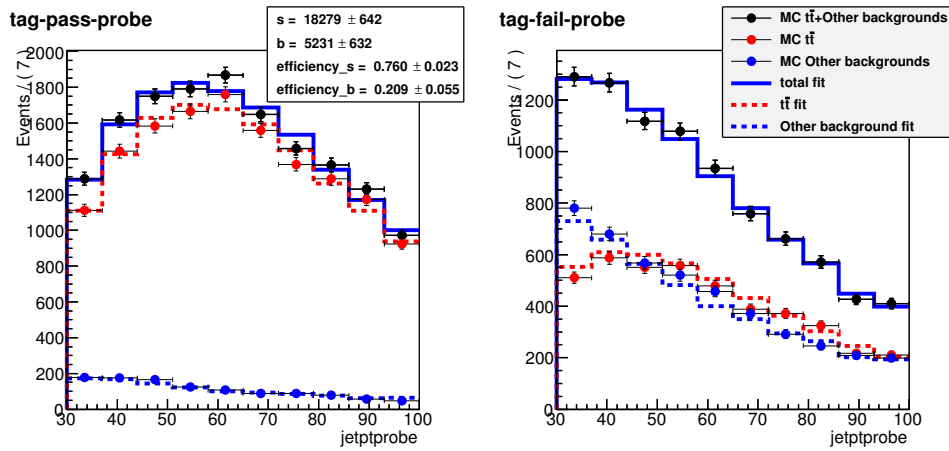


**Figure 4.6.:** Simultaneous fit of the *tag-pass-probe* and *tag-fail-probe* pairs in the MC.



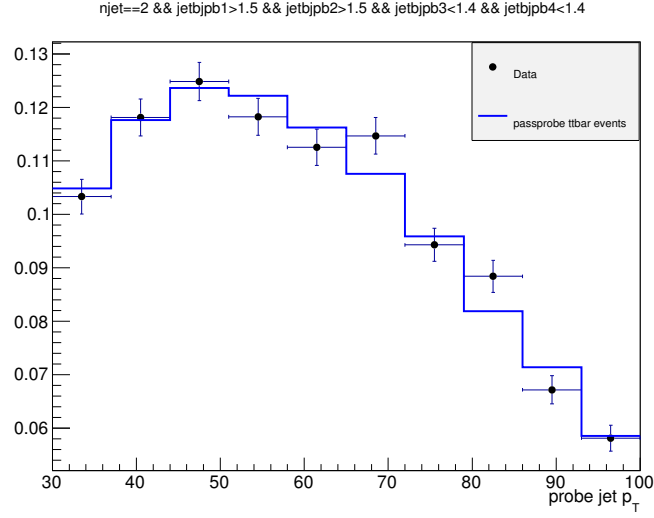
**Figure 4.7.:** Pulls of the  $\epsilon_s$  and  $\epsilon_b$  parameters in 5000 toy MC.

shape used to fit the true b-jets in the *tag-pass-probe* distribution. The result is shown in Fig. 4.9 and shows good agreement.



**Figure 4.8.:** Fit of a toy MC sample.



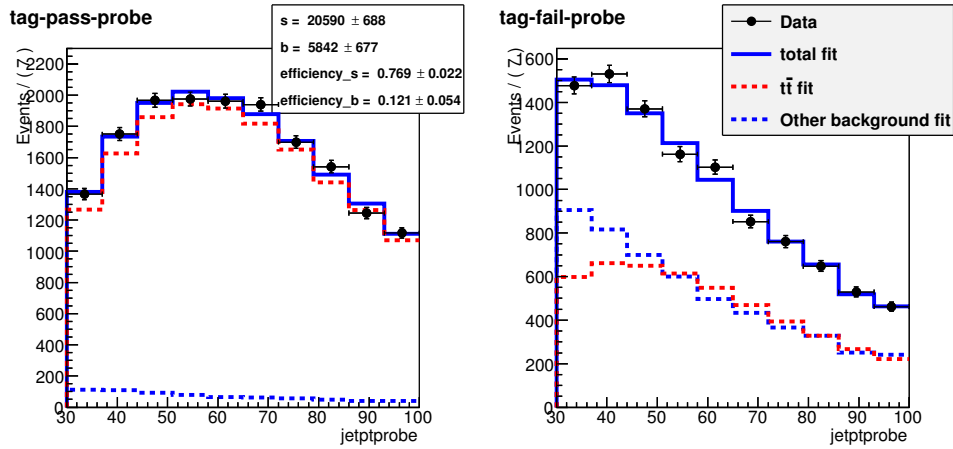


**Figure 4.9.:** Shape comparison for the *probe*  $p_T$  spectrum in data and in MC in a very pure  $t\bar{t}$  sample.

Finally the fit has been performed on data, as shown in Fig. 4.10, providing the following efficiencies:

$$\epsilon_s^{Data} = 0.769 \pm 0.022 \quad (4.12)$$

$$\epsilon_b^{Data} = 0.121 \pm 0.054 \quad (4.13)$$



**Figure 4.10.:** Simultaneous fit of the *tag-pass-probe* and *tag-fail-probe* pairs in data.

Further studies have been performed to assess the effect of the relative uncertainty on the  $t\bar{t}$  and  $tW$  event fractions. The same procedure described above has been applied to different simulation templates obtained varying the  $t\bar{t}$  and  $tW$  fractions within theoretical uncertainties, and the effect on the parameters extracted with the fit procedure is found to be well below the fit uncertainties.

### Data driven estimation

In addition to the b-tagging efficiency, the other ingredient to estimate the  $t\bar{t}$  background is the process cross section. The idea is to measure the cross section in a  $t\bar{t}$  enriched control region, that is called CtrlDD. CtrlDD is defined according to the lepton preselection cuts defined in Sec. 4.3.1, and requiring in addition at least one jet with *JetBProbability* score higher than 1.4.

From the simulation we derive the factor  $\alpha$  that connects CtrlDD to the signal region, calculating the ratio of  $t\bar{t}$  events in the two regions:

$$\alpha = \frac{N_{t\bar{t} \text{ MC}}^{SIG}}{N_{t\bar{t} \text{ MC}}^{CtrlDD}} \quad . \quad (4.14)$$

The number of events in the CtrlDD region in data is counted, subtracting the expected number of events from non- $t\bar{t}$  backgrounds, and obtaining  $N_{t\bar{t} \text{ Data}}^{CtrlDD}$ . Finally the number of expected  $t\bar{t}$  events in the signal region ( $N_{t\bar{t} \text{ Data}}^{SIG}$ ) is obtained as:

$$N_{t\bar{t} \text{ Data}}^{SIG} = \alpha N_{t\bar{t} \text{ Data}}^{CtrlDD} \quad . \quad (4.15)$$

In evaluating  $\alpha$  and its error the b-tagging efficiencies determined in Sec. 4.4.1 are used. For each event an efficiency scale factor and a mistag rate scale factor are derived, depending on whether the event falls in the signal or CtrlDD region.

$$SF_{SIG} = \left( \frac{1 - \epsilon_s^{Data}}{1 - \epsilon_s^{MC}} \right)^{\min(2, n_{b-jets})} \left( \frac{1 - \epsilon_b^{Data}}{1 - \epsilon_b^{MC}} \right)^{n_{non-b-jets}} \quad (4.16)$$

$$SF_{CtrlDD} = \left( \frac{\epsilon_s^{Data}}{\epsilon_s^{MC}} \right)^{(jet1 == b-jet)} \left( \frac{\epsilon_b^{Data}}{\epsilon_b^{MC}} \right)^{(jet1 == non-b-jets)} \quad (4.17)$$

where  $n_{b-jets}$  is the number of true b-jets in the event and  $n_{non-b-jets}$  is the number of non-b-jets in the event. The writing  $jet1 == b-jet$  ( $jet1 == non-b-jets$ ) is a boolean flag that is true when the leading jet, the one used for the CtrlDD selection, is (not) a true b-jet.

Since the efficiency and mistag rate that have been measured on data are close to the one in the simulation, it was decided to assume a scale factor of 1 for both b-tagging efficiency and mis-tag rate. This means that the central values of the scale factors defined in Eq. 4.16 and Eq. 4.17 is 1, but these numbers have an error that is derived assuming an uncertainty on  $\epsilon_s^{Data}$  and  $\epsilon_b^{Data}$  that covers both the statistical error from the fit of the two quantities and the difference with respect to the simulation. This results in an up and a down variations of the scale factors in the signal and CtrlDD regions, that is used to derive an error on  $\alpha$ .

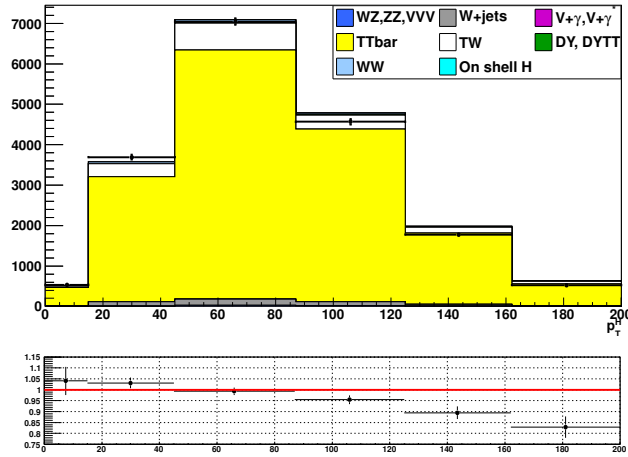
A data driven estimation of the top quark background with the method described above is performed in each of the  $p_T^H$  bins independently. The reason to make this estimation in  $p_T^H$  bins, rather than inclusively is explained in Fig. 4.11, where the  $p_T^H$  distribution is shown in the CtrlDD region normalized to the cross section measured by a specific analysis of CMS [REFERENZA](#). As shown in the ratio plot, an overall normalization factor would not be able to accommodate for the variations of the Data/simulation ratio from bin to bin.

The  $\alpha$  factors for each bin and the number of events in signal, CtrlDD regions in MC as well as in data are listed in Tab. 4.3.

A comparison of the  $m_{\ell\ell}$  distribution in the six  $p_T^H$  bins used in the analysis in CtrlDD after the data driven correction is shown in Fig. 4.12

$p_T^H$ bin	$N_{CTRL}^{DATA}$	$N_{CTRL}^{TOP}$	$N_{SIG}^{TOP}$	$\alpha$	$\Delta\alpha$
1	406.71	358.78	117.83	0.328	0.075
2	2930.14	2703.44	859.08	0.318	0.071
3	5481.02	5207.48	1506.05	0.289	0.065
4	4126.35	4032.56	861.22	0.214	0.052
5	1612.64	1654.27	304.69	0.184	0.055
6	647.50	760.37	201.70	0.265	0.147

**Table 4.3.:** Table of data driven scale factors.



**Figure 4.11.:**  $p_T^H$  distribution in the CtrlDD control region.

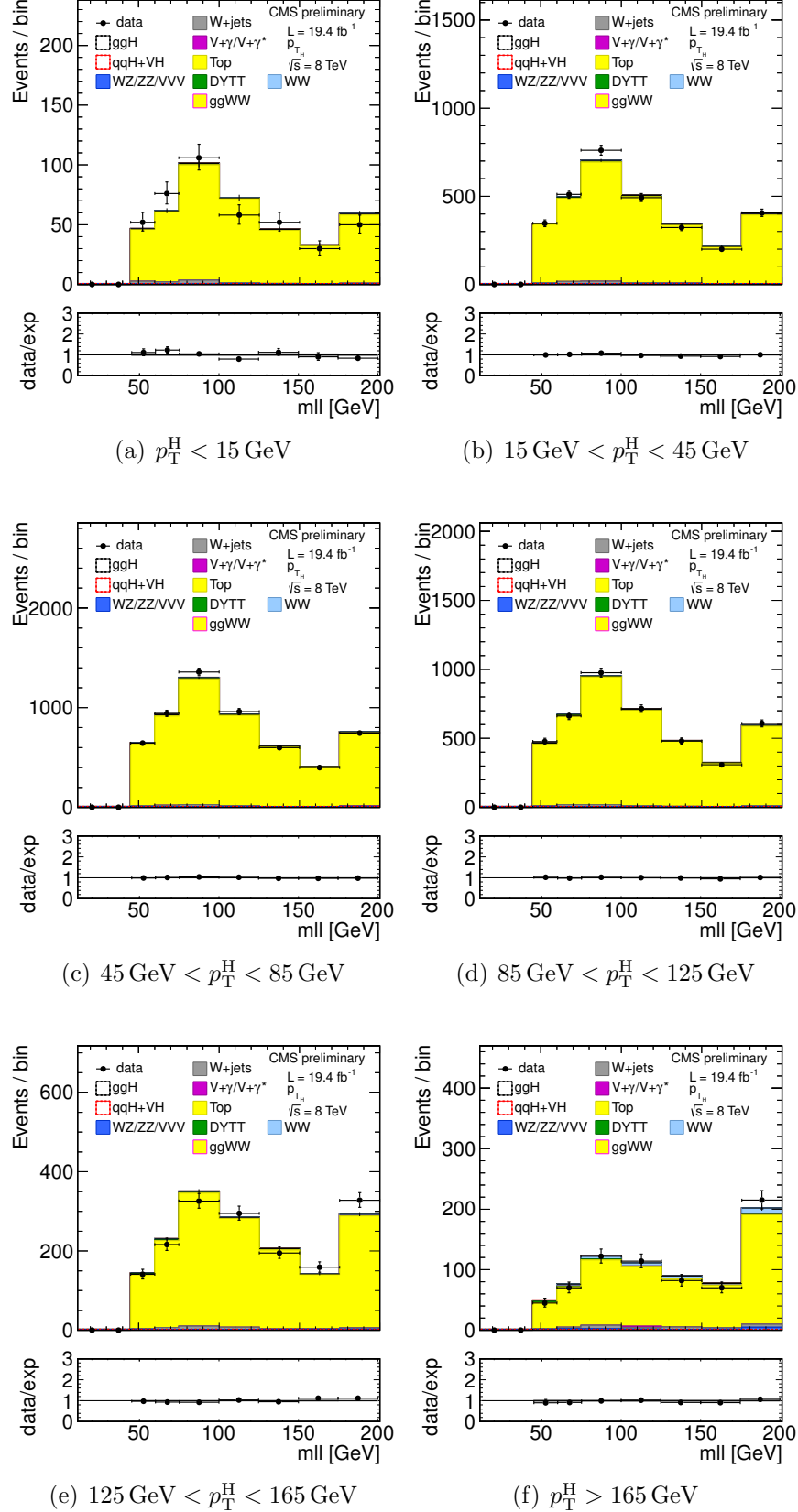


Figure 4.12.:  $m_{\ell\ell}$  distributions in the CtrlDD region for the different  $p_T^H$  bins.

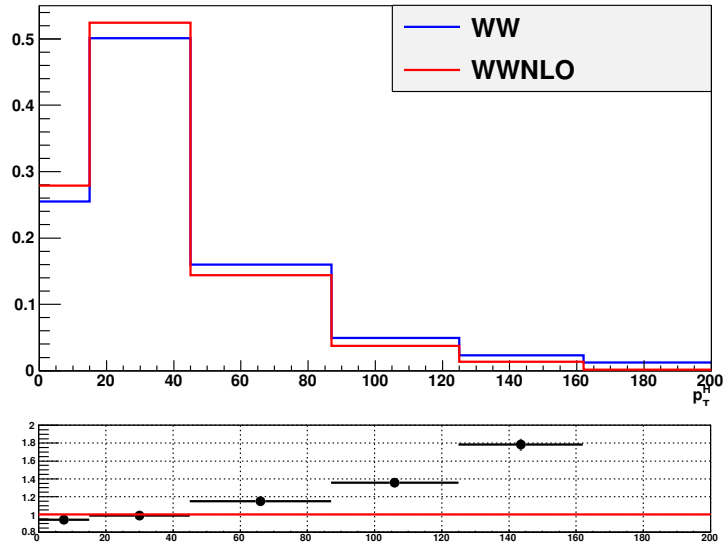
#### 4.4.2. WW background

For what the WW background shape is concerned, the prediction from the MC simulation has been used. This background is divided into six different parts, corresponding to the six  $p_T^H$  bins considered. In each bin the normalization of the WW background is left free to float, in such a way to adjust it to match the data during the fit procedure. In this way we minimize an effect that has been observed also in [29] **REFERENZA NON AN**, that is a difference in shape between the  $p_T^{WW}$  theory prediction and the distribution provided by the simulation, in our case by MADGRAPH generator.

In figure 4.13 a comparison is shown between the  $p_T^{WW}$  spectra of two different qqWW samples: one obtained with the MADGRAPH generator and the other after applying to the same distribution a reweighting in order to match the theoretical prediction at NLO+NNLL precision.

A shape discrepancy can be clearly observed and the effect becomes larger at high values of  $p_T^H$ .

In order to assess if these discrepancy has a not negligible effect on the shapes of the variables that we use for the fit,  $m_{\ell\ell}$  and  $m_T$ , we checked these distributions in every  $p_T^H$



**Figure 4.13.:** Comparison between the  $p_T^{WW}$  distributions obtained with two different MC generators: the blue line corresponds to the MADGRAPH generator and the red line refers to the same sample in which a reweighting has been applied in order to match the theoretical prediction at NLO+NNLL precision.

bin, comparing several samples. In particular we compared the MADGRAPH sample used for the nominal shape, the MADGRAPH sample with NLO+NNLL reweighting, a POWHEG NLO sample and an AMC@NLO sample. The results of this comparison are shown in figures 4.14 and 4.15. The discrepancy in shape among the different models is within the statistical accuracy of the MC samples.

### 4.4.3. Other backgrounds

#### W+jets background

Backgrounds containing one or two fake leptons are estimated from events selected with relaxed lepton quality criteria, using the efficiencies for real and fake leptons to pass the tight lepton quality cuts of the analysis.

A data-driven approach, described in detail in [30] and [31] [REFERENCE](#), is pursued to estimate this background. A set of loosely selected lepton-like objects, referred to as the 'fakeable object' or "denominator" from here on, is defined in a sample of events dominated by dijet production. The denominator object definition used in the full 2012 data is described in [32] [REFERENZA](#).

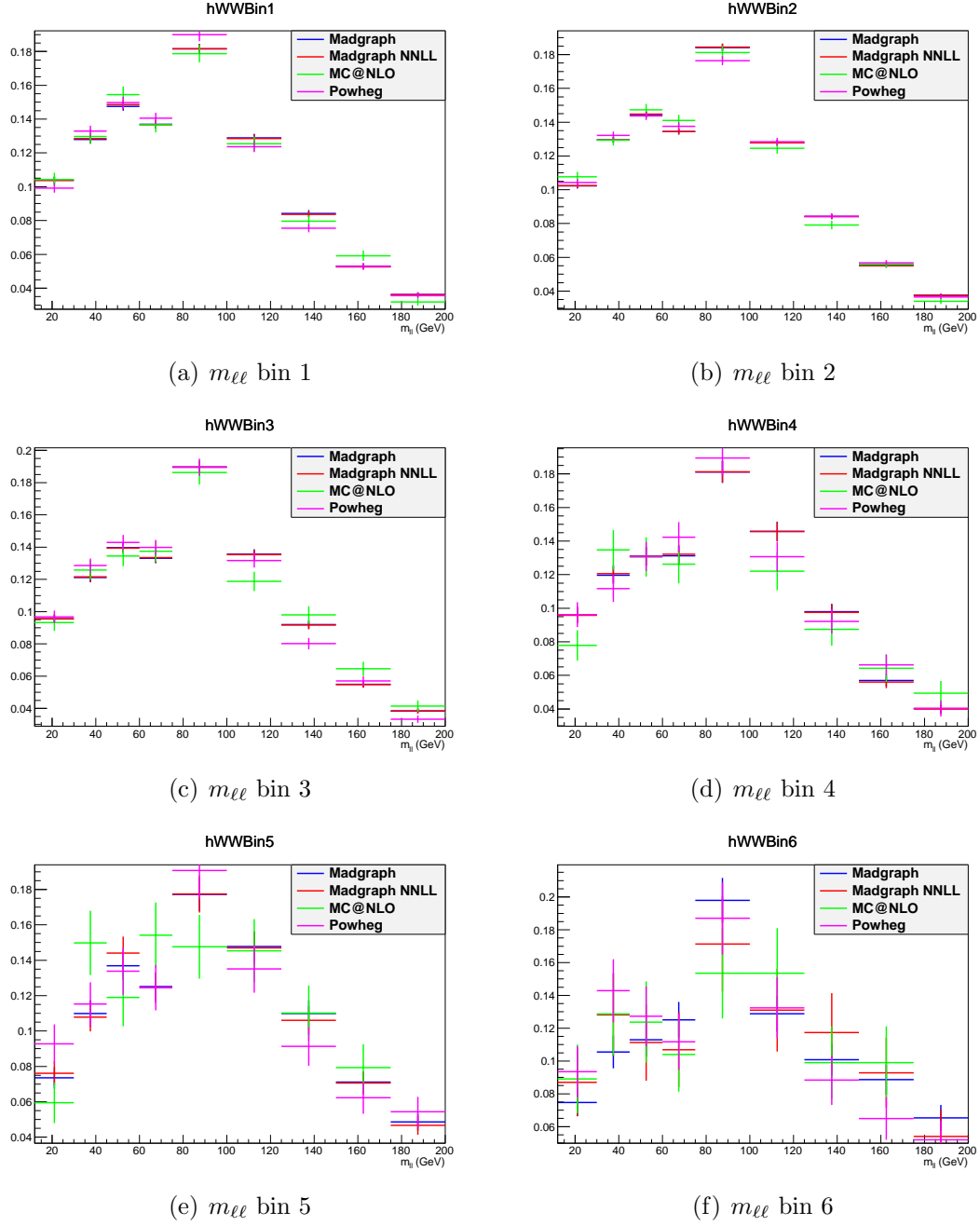
To measure the fake rate we count how many fakeable objects pass the full lepton selection of the analysis, parameterized as a function of the phase space of the fakeable lepton, therefore it is extracted in bins of  $\eta$  and  $p_T$ .

The ratio of the fully identified lepton, referred as "numerator", to the fakeable objects is taken as the probability for a fakeable object to fake a lepton:

$$Fake\ Rate = \frac{\#of\ fully\ reconstructed\ leptons}{\#of\ fakeable\ objects} \quad (4.18)$$

It is then used to extrapolate from the loose leptons sample to a sample of leptons satisfying the full selection.

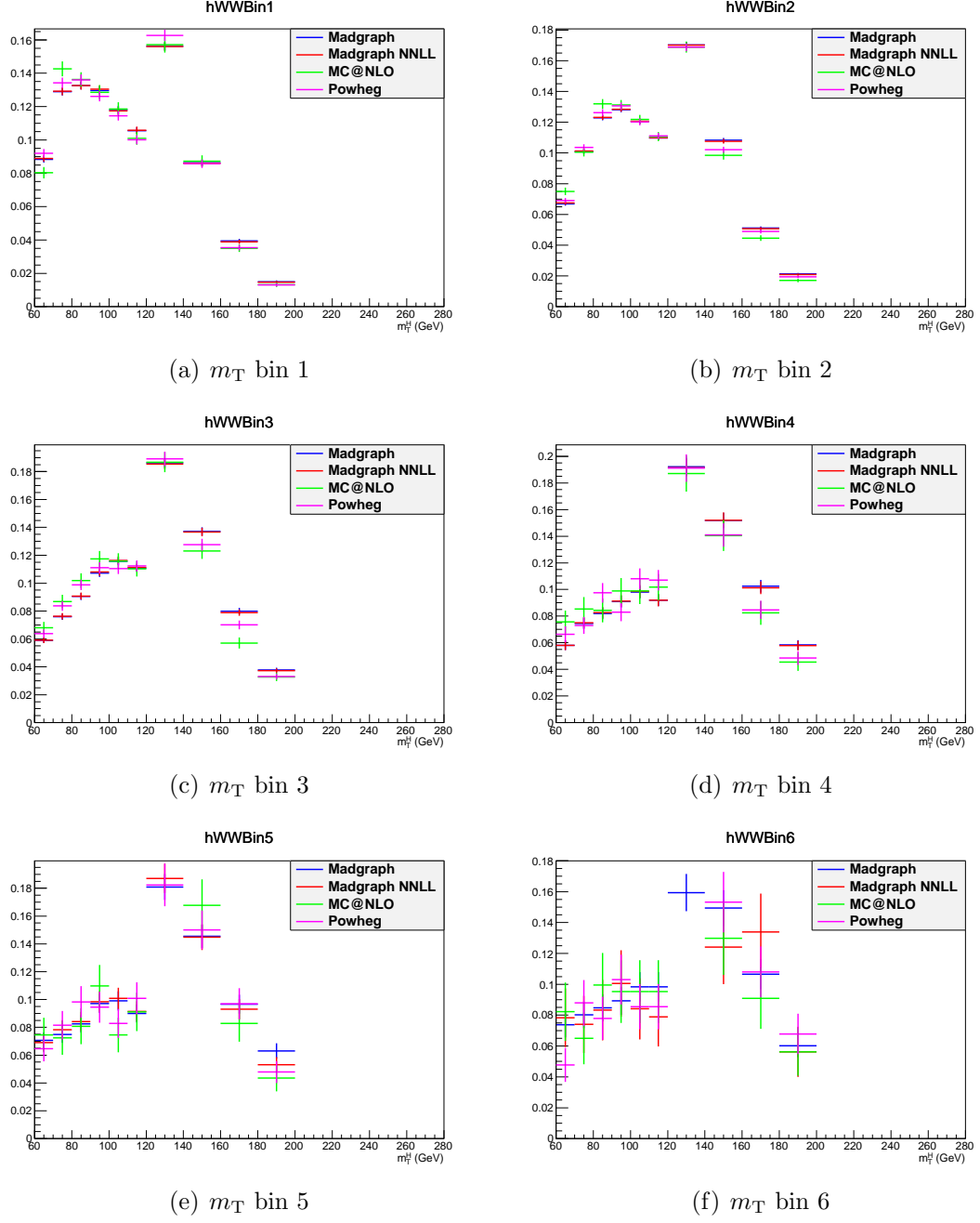
The details of the method implementation can be found in [20] [REFERENCE](#). The systematic uncertainty is evaluated by varying the jet thresholds in the di-jet control sample,



**Figure 4.14.:** Comparison between the default WW background sample and other theoretical models for the  $m_{\ell\ell}$  distributions in every  $p_T^H$  bin.

and by performing a closure test in the same-sign data sample (see [20]) **REFERENZE**. In both cases it is about 36%.





**Figure 4.15.:** Comparison between the default WW background sample and other theoretical models for the  $m_T$  distributions in every  $p_T^H$  bin.

## Drell-Yan to $\tau\tau$ background

The low  $E_{\text{T}}^{\text{miss}}$  threshold in  $e\mu$  final state requires the consideration of the contribution from  $Z/\gamma^* \rightarrow \tau^+\tau^-$  that is in fact estimated from data. This is accomplished by using  $Z/\gamma^* \rightarrow \mu^+\mu^-$ -events and replacing muons with a simulated  $\tau \rightarrow l\nu_\tau\bar{\nu}_e$  decay [33] [REFERENCE](#). After replacing muons from  $Z/\gamma^* \rightarrow \mu^+\mu^-$ -decays with simulated  $\tau$  decays, the set of pseudo  $Z/\gamma^* \rightarrow \tau^+\tau^-$ -events undergoes the reconstruction step.

Good agreement in kinematic distributions for this sample and a MC based  $Z/\gamma^* \rightarrow \tau^+\tau^-$ -sample is found. The global normalization of pseudo  $Z/\gamma^* \rightarrow \tau^+\tau^-$ -events is checked in the low  $m_{\text{T}}$  spectrum where a rather pure  $Z/\gamma^* \rightarrow \tau^+\tau^-$ -sample is expected.

## ZZ, WZ and $W\gamma$ backgrounds

The WZ and ZZ backgrounds are partially estimated from data when the two selected leptons come from the same Z boson. If the leptons come from different bosons the contribution is expected to be small. The WZ component is largely rejected by requiring only two high  $p_{\text{T}}$  isolated leptons in the event.

The  $W+\gamma^{(*)}$  background, where the photon decays to an electron-positron pair, is expected to be very small, thanks to the stringent photon conversion requirements. Since the WZ simulated sample has a generation level cut on the di-lepton invariant mass ( $m_{\ell\ell} > 12$  GeV) and the cross-section raises quickly with the lowering of this threshold, a dedicated MADGRAPH sample has been produced with lower momentum cuts on two of the three leptons ( $p_{\text{T}} > 5$  GeV) and no cut on the third one. The surviving contribution estimated with this sample is still very small, and since the uncertainty on the cross-section for the covered phase space is large, a conservative 100% uncertainty has been given to it. A  $k$ -factor for  $W+\gamma^*$  of  $1.5 \pm 0.5$  based on a dedicated measurement of tri-lepton decays,  $W+\gamma^* \rightarrow e\mu\mu$  and  $W+\gamma^* \rightarrow \mu\mu\mu$ , is applied [34] [REFERENCE](#). The contribution of  $W+\gamma^{(*)}$  is also constrained by a closure test with same sign leptons on data, which reveals a good compatibility of the data with the expected background.

## 4.5. Systematic uncertainties

Systematic uncertainties play an important role in this analysis where no strong mass peak is expected due to the presence of undetected neutrinos in the final state. One of the most important sources of systematic uncertainty is the normalization of the backgrounds that are estimated on data control samples whenever is possible.

### 4.5.1. Background normalization uncertainties

The signal extraction is performed subtracting the estimated backgrounds to the event counts in data. This uncertainty depends on the background:

- **$t\bar{t}$  and  $tW$  backgrounds:** The efficiency on jets b-tagging is estimated using the Tag&Probe technique in data and Monte Carlo control regions, as explained in 4.4.1. A per-jet scale factor, which takes into account the possibly different efficiency of the anti b-tagging selection in data and MC, is computed by means of the efficiency measured with the Tag&Probe method. The Tag&Probe method has been used also to measure the mistag rates in data and MC, which are the probability to b-tag a jet that is not produced by the hadronization of a b quark. These factors are used to reweigh the Top MC samples as explained in 4.4.1.

The uncertainties provided by the Tag&Probe fit are then propagated to the factor  $\alpha$  that is used in the top data driven estimation 4.4.1. These uncertainties are embedded in a systematic error that affects the shape of the Top background in each  $p_T^H$  bin. In fact the Top background has been splitted in six different contributions, one for each bin of  $p_T^H$ , and a different uncertainty has been associated to each background as well.

Provided that our Top MC samples include both  $t\bar{t}$  and  $tW$  processes, a systematic uncertainty related to the  $tW/t\bar{t}$  fraction has been included. In fact, a relative variation of the contribution of these two processes could modify the shape of the Top MC sample, and is thus included as a shape uncertainty affecting the Top shape in each bin of  $p_T^H$  in a correlated way.

- **$W$ +jets background:** It is estimated with data control sample as described in Sec.4.4.3. With  $19.4\text{fb}^{-1}$  at 8 TeV, the uncertainty receives similar contributions from statistics and systematic error (mainly jet composition differences between the fake

rate estimation sample and the application sample), the total error being about 40%, dominated by the closure test of the method on Monte Carlo [20].

- **WZ,ZZ,W $\gamma^{(*)}$  backgrounds:** those backgrounds, which are expected to give a small contribution, are estimated from simulation. We assign the uncertainties on cross sections reported in [35, 36]: 4% to WZ, 2.5% to ZZ. We also assign 30% on W $\gamma$  [37] and 30% on W $\gamma^{(*)}$  according to the uncertainty on the normalization study (see section 4.4.3).

#### 4.5.2. Experimental uncertainties

The following experimental systematic sources have been taken into account:

- **Luminosity:** Using the online luminosity monitoring CMS reached an uncertainty on the luminosity of 2.6% at 8 TeV.
- **Trigger efficiency.** The uncertainties for both electrons and muons are at 1-2% level, which is added together to the lepton efficiency uncertainty.
- **Lepton reconstruction and identification efficiency:** The lepton reconstruction and identification efficiencies are measured with the Tag&Probe method in data. To correct for the difference in the lepton identification efficiencies between data and MC, a scale factor is applied to MC. The uncertainties resulting from this procedure on the lepton efficiencies are 4% for electrons and 3% for muons.
- **Muon momentum and electron energy scale:** The momentum scale of leptons have relatively large uncertainties due to different detector effects. For electrons a scale uncertainty of 2% for the barrel, and 4% for the endcaps respectively, is assigned. For muons, a momentum scale uncertainty of 1.5%, independent of its pseudorapidity, is assigned.
- **$E_T^{\text{miss}}$  modeling:** The  $E_T^{\text{miss}}$  measurement is affected by the possible mis-measurement of individual particles addressed above, as well as the additional contributions from the pile-up interactions. The effect of the missing transverse momentum resolution on the event selection is studied by applying a Gaussian smearing of 10% on the  $x$ - and  $y$ -components of the missing transverse momentum. All correlated variables, like the transverse mass, are recalculated.

- 
- **Jet energy scale (JES) uncertainties:** It affects both the jet multiplicity and the jet kinematic variables, such as  $m_{jj}$ . We estimate this uncertainty applying variations of the official jet uncertainties on the JES (which depend on  $\eta$  and  $p_T$  of the jet [38]) and compute the variation of the selection efficiency.
  - **B-mistag modeling.** A fraction of signal events is rejected because erroneously identified as b-jet by the *JetProbability* tagger. The mistag rate comes with an uncertainty due to different modeling of the b-tagging performance in data and MC. The mistag rate has been measured in data and MC with a Tag&Probe technique, as described in Sec. 4.4.1, and a per jet scale factor has been derived from the ratio of the mistag rates measured in data and MC. This scale factor is consistent with one, as shown in Sec. 4.4.1. The scale factor and its uncertainty are used to reweight each signal event with a weight corresponding to the number of non-b-jets in the event.
  - **Pileup multiplicity:** Some of the variables used in the analysis are affected by the average number of pileup interactions. The simulated events have been reweighted according the instantaneous luminosity measured on data. The error in the average number of pileup interactions measured in data and the simulation of the modeling and physics aspects of the pileup simulation gives an uncertainty of 5% on the distribution used in the reweighting procedure. This uncertainty is propagated through all the analysis, and the estimated uncertainty on the efficiency is 2%.

### 4.5.3. Theoretical uncertainties

- **QCD scale uncertainties:** The uncertainties on the total cross sections due to the choice of the renormalization and factorization scale are assigned to MC-driven backgrounds (ggWW, WZ, ZZ). For the signal processes these uncertainties are separated in two categories: those affecting the selection efficiency and those affecting the jet bin fractions. The effect of renormalization and factorization scale on the selection efficiency is of the order of 2% for all processes. Although this analysis is inclusive in number of jets, the effect of the QCD scale variation on the jet bin migrations has to be taken into account because of the b-tagging veto efficiency. The efficiency of this selection depends on jet multiplicity and the effect of the QCD scale variation has been evaluated using the Stewart-Tackman method, as explained in 4.5.3.

- **PDFs uncertainties:** The utilization of different PDF sets can affect both the normalization and the shapes of the signal contributions. The uncertainty related due to the variations in the choice of PDFs is considered following the PDF4LHC [39, 40] prescription, using CT10, NNPDF2.1 [41] and MSTW2008 [42] PDF sets.
- **WW:** Due to the fact that the WW shape is entirely taken from simulation, the analysis is strongly relying on theoretical models and can thus be strongly affected by their uncertainties. Especially higher order QCD radiative effects have an influence on the generated WW shape. To study this impact, the shapes of the distributions produced with the MADGRAPH generator (which is the generator for the MC simulation used in the analysis) are compared to the ones produced with MC@NLO. The comparison is performed separately in each bin of  $p_T^H$  and the uncertainty includes shape differences originating from the renormalization and factorization scale choice. A comparison of the  $m_{\ell\ell}$  and  $m_T$  shapes for the WW background using different MC generators is reported in section 4.4.2.

### Jets multiplicity uncertainty

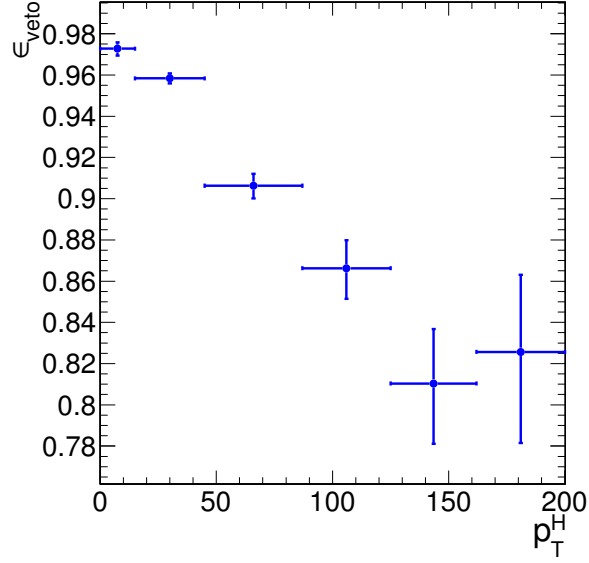
The jet bin uncertainty on ggH sample has been evaluated using the Stewart-Tackman method, following the recipe proposed in “Procedure for the LHC Higgs boson search combination” [43].

Three nuisance parameters have been calculated according to the table 4.4, where  $\kappa = \sqrt{\exp(\epsilon_-)\exp(\epsilon_+)}$  and  $\epsilon_{\pm}$  are relative QCD scale uncertainties. Exclusive cross sections for 0, 1 and 2-jet bins are calculated for the default QCD scale and their variation by changing the scale by a factor of 2 and 1/2 (up/down). The  $f_n$  constants represent the exclusive theoretical  $n$  jet bin fractions.

In this analysis, which is inclusive in number of jets, we have to include the jet binning uncertainties only if the b-tagging veto efficiency depends on the number of jets in the event. The veto efficiency has been calculated in all the  $p_T^H$  bins defined in the analysis and as a function of jets multiplicity. The results are shown in figures 4.16 and 4.17. The drop of the veto efficiency at high values of the Higgs  $p_T$  is due to the relation with jets multiplicity.

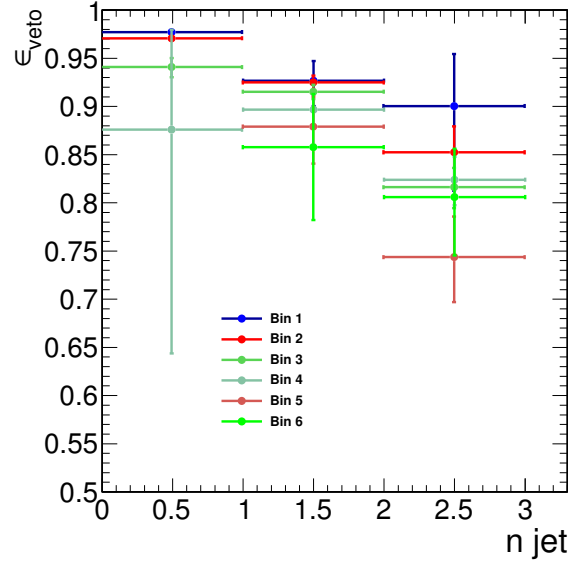
**Table 4.4.:** Numerical calculation for the systematics uncertainty of jet binning.

Nuisance parameter	0-jet bin	1-jet bin	2-jet bin
QCDscale	$\kappa = (\kappa_{\geq 0})^{\frac{1}{f_0}}$		
QCDscale1in	$\kappa = (\kappa_{\geq 1})^{-\frac{f_1+f_2}{f_0}}$	$\kappa = (\kappa_{\geq 1})^{\frac{f_1+f_2}{f_1}}$	
QCDscale2in		$\kappa = (\kappa_{\geq 2})^{-\frac{f_2}{f_1}}$	$\kappa = (\kappa_{\geq 2})$

**Figure 4.16.:** Efficiency of the b-tagging veto in different bins of  $p_T^H$ .

The nuisance parameters reported in table 4.4 have then been calculated for each  $p_T^H$  bin embedding the veto efficiency and using the following formulas:

$$QCDscale_{ggH} = \frac{ggH0 * f_0 * \epsilon_0 + ggH1in0 * f_1 * \epsilon_1}{ggH0 * f_0 * \epsilon_0 + ggH1in0 * f_1 * \epsilon_0} \quad (4.19)$$



**Figure 4.17.:** Efficiency of the b-tagging veto in different bins of  $p_T^H$ , as a function of number of jets.

$$QCDscale\_ggH1in = \frac{ggH1in1 * f_1 * \epsilon_1 + ggH2in1 * f_2 * \epsilon_2}{ggH1in1 * f_1 * \epsilon_1 + ggH2in1 * f_2 * \epsilon_1} \quad (4.20)$$

$$QCDscale\_ggH2in = 1 \quad (4.21)$$

These nuisance parameters are expected to be equal to one in case the efficiency is independent on the number of jets, i.e if  $\epsilon_0 = \epsilon_1 = \epsilon_2$ .

The values obtained are reported in table 4.5 divided in bins of  $p_T^H$ .

**Table 4.5.:** Values of the jet binning nuisance parameters for different  $p_T^H$  bins.

	Bin 1	Bin 2	Bin 3	Bin 4	Bin 5	Bin 6
QCDscale_ggH	0.998	0.993	0.989	1.000	1.000	1.000
QCDscale_ggH1in	0.997	0.993	0.984	0.975	0.946	0.974



#### 4.5.4. Monte Carlo statistics

Due to the large range of weights to correct other MC pile up distribution to match that in data, the effective size of the MC samples are sometimes smaller than the actual number of events in the sample. The statistical uncertainty of the event yields estimated from MC samples is reflected in the final result.

#### 4.5.5. Treatment of systematics in the shape analysis

One can distinguish between normalization uncertainties, where a systematic effect is changing the normalization assuming the shape is not affected, and shape uncertainties where the actual change in the shape of the distribution is taken into account. The normalization uncertainties enter the shape analysis as a constant normalization factor, whereas for shape uncertainties the nominal and the  $+1\sigma$  and  $-1\sigma$  shapes enter the analysis in form of three histograms with the same normalization.

For the  $W$ +jets background, the shape differences for different jet  $p_T$  thresholds in the di-jet control sample are considered separately for electron and muon fakes, while the other sources of systematics are taken as normalization uncertainties as in the cut-based analysis.

Effects from experimental uncertainties are studied by applying a scaling and/or smearing of certain variables of the physics objects, followed by a subsequent recalculation of all the correlated variables. This is done for Monte Carlo simulation, to account for possible systematic mismeasurements of the data. All experimental sources from Section 4.5.2 but luminosity are treated both as normalization and shape uncertainties. For background with a data-driven normalization estimation, only the shape uncertainty is considered.

To account for statistical uncertainties, for each distribution going into the shape analysis, the  $+1\sigma$  and  $-1\sigma$  shapes were obtained by adding/subtracting the statistical error in each bin and renormalize it to the nominal distribution. In addition to this procedure a constant normalization uncertainty due to the finite statistics of the sample, used to extract the shape, is assigned.

A summary of the main sources of systematic uncertainty and the corresponding estimate is reported in Table 4.6.

**Table 4.6.:** Main sources of systematic uncertainties and their estimate. The first category reports the uncertainties in the normalization of background contributions. The experimental and theoretical uncertainties refer to the effect on signal yields. A range is specified if the uncertainty varies across the  $p_T^H$  bins.

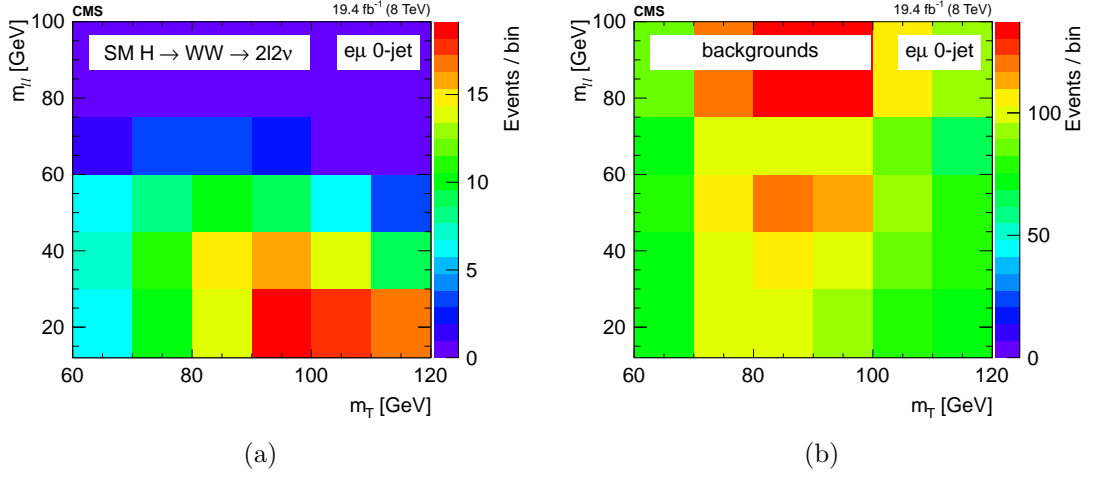
<b>Uncertainties in backgrounds contributions</b>	
Source	Uncertainty
$t\bar{t}$ , $tW$	20–50%
$W$ + jets	40%
$WZ$ , $ZZ$	4%
$W\gamma^{(*)}$	30%
<b>Effect of the experimental uncertainties on the signal and background yields</b>	
Source	Uncertainty
Integrated luminosity	2.6%
Trigger efficiency	1–2%
Lepton reconstruction and identification	3–4%
Lepton energy scale	2–4%
$E_T^{\text{miss}}$ modelling	2%
Jet energy scale	10%
Pileup multiplicity	2%
$b$ mistag modelling	3%
<b>Effect of the theoretical uncertainties on signal yield</b>	
Source	Uncertainty
$b$ jet veto scale factor	1–2%
PDF	1%
$WW$ background shape	1%

## 4.6. Signal extraction

According to the “blinding” policy of the CMS Collaboration, the strategy of the analysis has been scrutinized and approved by a selected committee of internal reviewers before looking at the data in the signal region. This approach prevents the analysts from being biased by the data in the developing phase of the analysis. Below are shown the results after having looked at the data.

### 4.6.1. Fitting procedure

The signal, including ggH, VBF, and VH production mechanisms, is extracted in each bin of  $p_T^H$  by performing a binned maximum likelihood fit simultaneously in all  $p_T^H$  bins to a two-dimensional template for signals and backgrounds in the  $m_{\ell\ell}$ - $m_T$  plane. The variables used for the two-dimensional template are chosen for their power to discriminate signal and background contributions. This is shown in Fig. 4.18, where the two-dimensional MC distributions are shown for the signal and background processes in the 0-jets category.



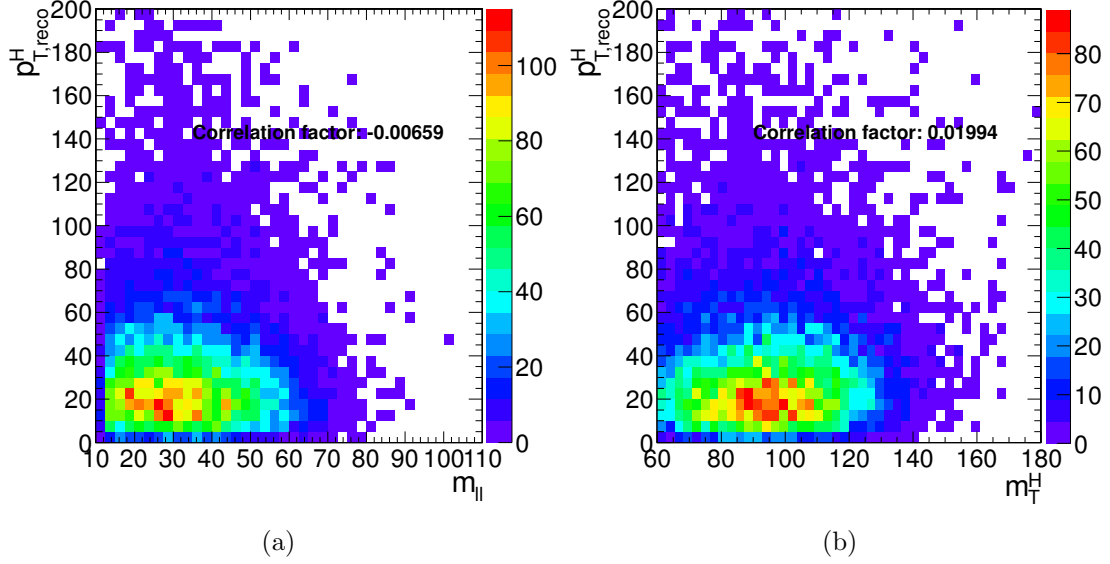
**Figure 4.18.:** Two-dimensional  $m_{\ell\ell}$ - $m_T$  distribution for signal (a) and background (b) processes in the 0-jets category.

Six different signal strength parameters are extracted from the fit, one for each  $p_T^H$  bin. The relative contributions of the different Higgs production mechanisms in the signal template are taken to be the same as in the SM. The systematic uncertainty sources are considered as nuisance parameters in the fit.

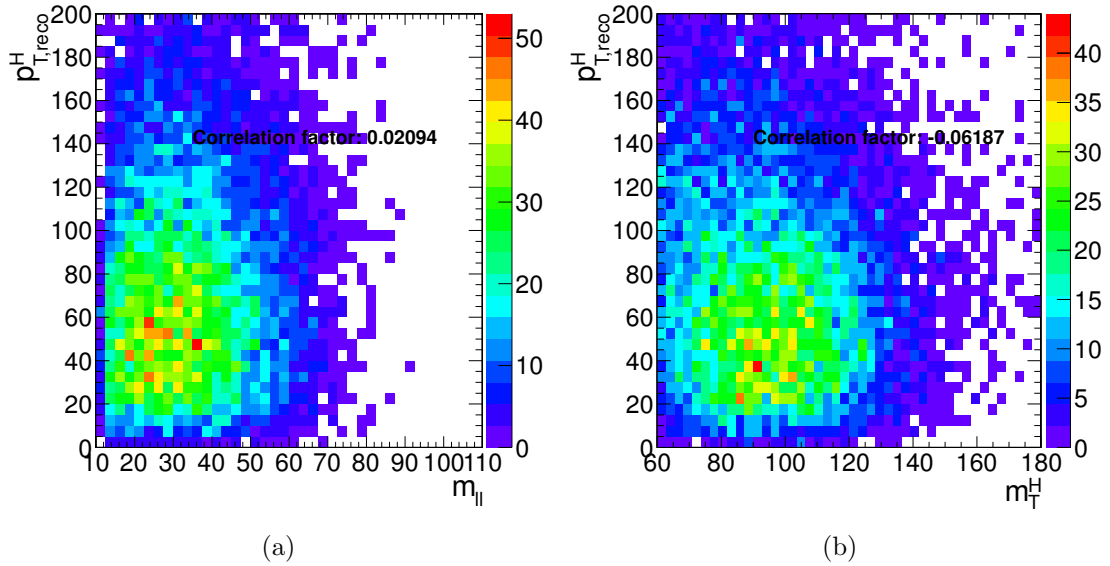
The binning of the  $m_{\ell\ell}$  and  $m_T$  templates is chosen to be:

- $m_{\ell\ell}$ : [12, 30, 45, 60, 75, 100, 125, 150, 175, 200]
- $m_T$ : [60, 70, 80, 90, 100, 110, 120, 140, 160, 180, 200, 220, 240, 280]

To avoid a dependence of the results on the variables used for the template fit,  $m_{\ell\ell}$  and  $m_T$  need to be uncorrelated with respect to  $p_T^H$ . This has been verified and the correlation between the discriminating variables and  $p_T^H$  is shown in Fig. 4.19 and Fig. 4.20 for ggH and VBF production modes respectively.



**Figure 4.19.:** Correlation between  $p_T^H$  and  $m_{\ell\ell}$  (a) and between  $p_T^H$  and  $m_T$  (b) after the full selection for the ggH production mode.



**Figure 4.20.:** Correlation between  $p_T^H$  and  $m_{\ell\ell}$  (a) and between  $p_T^H$  and  $m_T$  (b) after the full selection for the VBF production mode.

The relative contribution for different production mechanisms in the input signal template is taken to be the same as the SM. The signal strength  $\mu$  in each bin, i.e. the ratio between the measured cross section and the SM one,  $\mu = \sigma/\sigma_{\text{SM}}$ , is allowed to float between -10 and

+10, thus allowing negative values. This is mainly intended to allow the error bars to float below 0.

Because of detector resolution effects, some of the reconstructed  $H \rightarrow WW$  signal events might originate from outside the fiducial phase space. These out-of-fiducial signal events cannot be precisely handled by the unfolding procedure and must be subtracted from the measured spectrum. The  $p_T^H$  distribution of the out-of-fiducial signal events is taken from simulation, and each bin is multiplied by the corresponding measured signal strength before performing the subtraction.

At the end, the number of events in each bin  $i$  of the measured spectrum is:

$$N_i = \mu_i(s_i - f_i) \quad , \quad (4.22)$$

where  $s_i$  and  $f_i$  are respectively the number of signal and fake events expected from simulation and  $\mu_i$  is the measured signal strength.

The fit makes use of the binned maximum likelihood approach. Each source of systematic uncertainty is represented by a nuisance parameter in the likelihood function. **add some techincality about the fit, i.e. different priors of the nuisance parameters etc.**

Before running the fit on the data, the same procedure has been applied on the so called *Asimov data set*<sup>1</sup>, which provides a simple method to obtain the signal sensitivity before looking at the data [44].

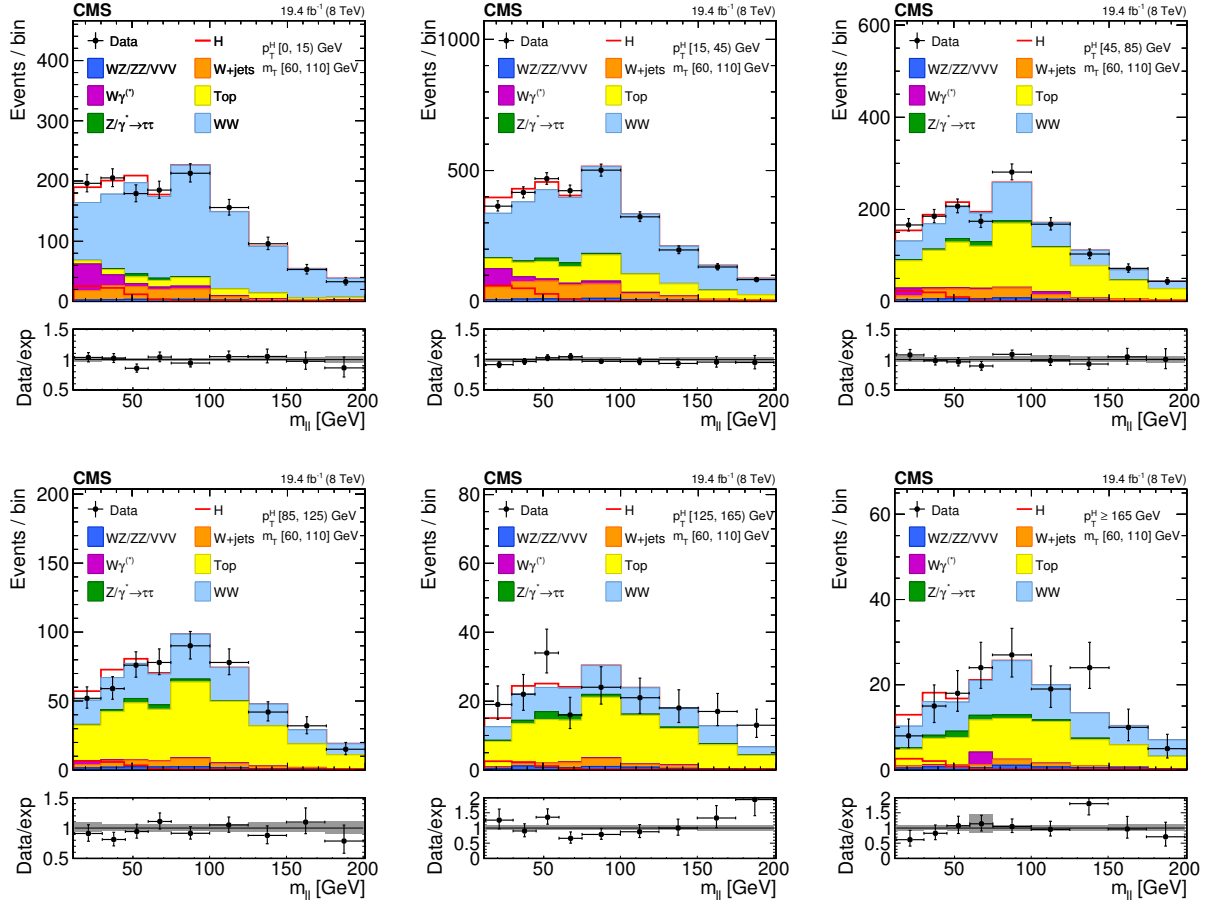
### 4.6.2. Signal and background yields

A comparison of data and background prediction is shown in Fig. 4.21, where the  $m_{\ell\ell}$  distribution is shown for the six  $p_T^H$  bins. Distributions correspond to the  $m_T$  window of [60, 110] GeV, in order to emphasize the signal contribution [18]. The  $m_T$  distributions are shown in Fig. 4.22 and correspond to the  $m_{\ell\ell}$  window of [12, 75] GeV.

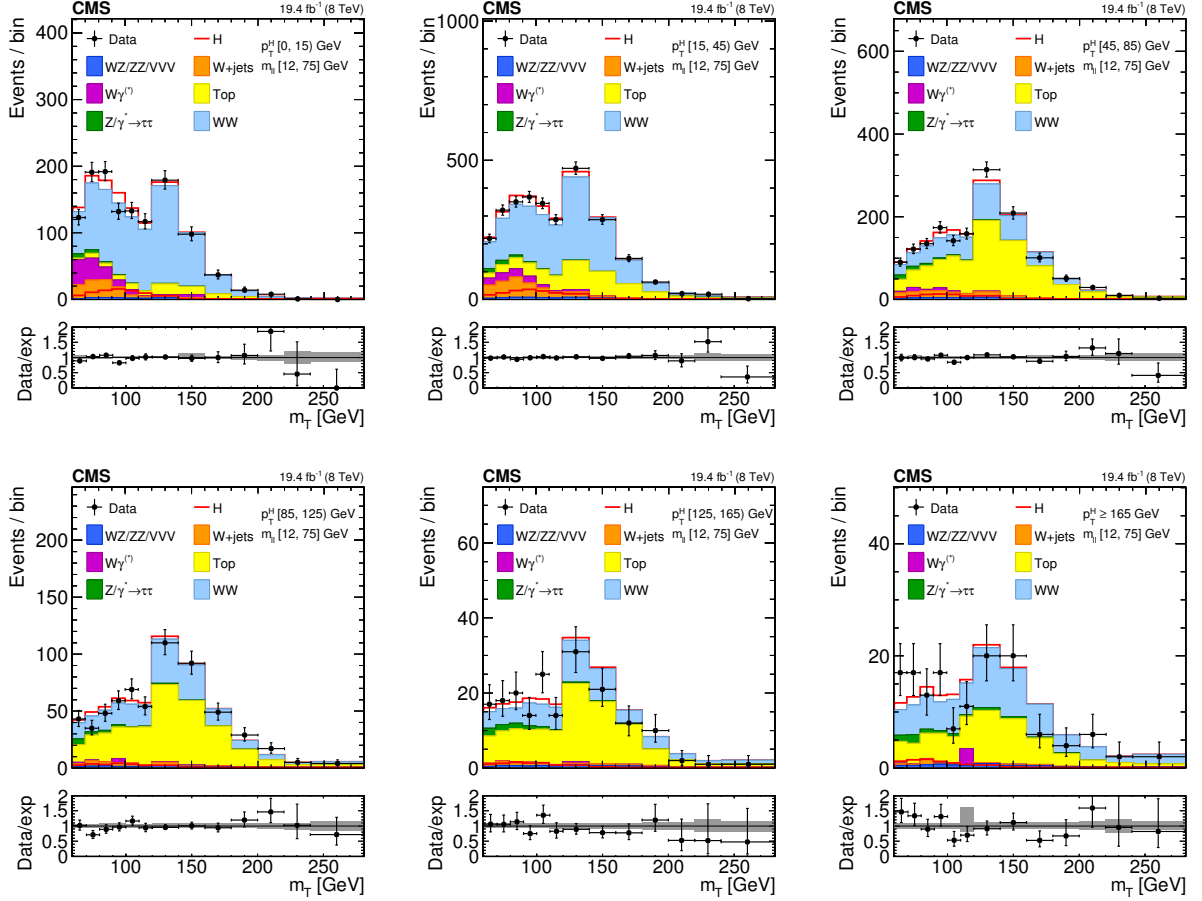
The signal and background yields after the analysis selection are reported in Table 4.7.

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<sup>1</sup>In a parallel reality imagined by the science fiction writer I. Asimov, politics was run in a peculiar way: instead of mobilizing millions of people to cast their vote to deliberate on something, an algorithm was used to select an individual “average” person, and then this person was asked to take the decision on that matter.



**Figure 4.21.:** Distributions of the  $m_{\ell\ell}$  variable in each of the six  $p_T^H$  bins. Background normalizations correspond to the values obtained from the fit. Signal normalization is fixed to the SM expectation. The distributions are shown in an  $m_T$  window of [60,110] GeV in order to emphasize the Higgs boson (H) signal. The signal contribution is shown both stacked on top of the background and superimposed to it. Ratios of the expected and observed event yields in individual bins are shown in the panels below the plots. The uncertainty band shown in the ratio plot corresponds to the envelope of systematic uncertainties after performing the fit to the data.



**Figure 4.22.:** Distributions of the  $m_T$  variable in each of the six  $p_T^H$  bins. Background normalizations correspond to the values obtained from the fit. Signal normalization is fixed to the SM expectation. The distributions are shown in an  $m_{\ell\ell}$  window of [12,75] GeV in order to emphasize the Higgs boson (H) signal. The signal contribution is shown both stacked on top of the background and superimposed to it. Ratios of the expected and observed event yields in individual bins are shown in the panels below the plots. The uncertainty band shown in the ratio plot corresponds to the envelope of systematic uncertainties after performing the fit to the data.

**Table 4.7.:** Signal prediction, background estimates and observed number of events in data are shown in each  $p_T^H$  bin for the signal after applying the analysis selection requirements. The total uncertainty on the number of events is reported. For signal processes, the yield related to the ggH are shown, separated with respect to the contribution of the other production mechanisms (XH=VBF+VH). The WW process includes both quark and gluon induced contribution, while the Top process takes into account both  $t\bar{t}$  and  $tW$ .

$p_T^H$ [GeV]	0-15	15-45	45-85	85-125	125-165	165- $\infty$
ggH	$73 \pm 3$	$175 \pm 5$	$59 \pm 3$	$15 \pm 2$	$5.1 \pm 1.5$	$4.9 \pm 1.4$
XH=VBF+VH	$4 \pm 2$	$15 \pm 4$	$16 \pm 4$	$8 \pm 2$	$3.8 \pm 1.1$	$3.0 \pm 0.8$
Out-of-fiducial	$9.2 \pm 0.5$	$19.9 \pm 0.7$	$11.4 \pm 0.6$	$4.4 \pm 0.3$	$1.6 \pm 0.2$	$2.4 \pm 0.2$
Data	2182	5305	3042	1263	431	343
Total background	$2124 \pm 128$	$5170 \pm 321$	$2947 \pm 293$	$1266 \pm 175$	$420 \pm 80$	$336 \pm 74$
WW	$1616 \pm 107$	$3172 \pm 249$	$865 \pm 217$	$421 \pm 120$	$125 \pm 60$	$161 \pm 54$
Top	$184 \pm 38$	$1199 \pm 165$	$1741 \pm 192$	$735 \pm 125$	$243 \pm 51$	$139 \pm 49$
W+jets	$134 \pm 5$	$455 \pm 10$	$174 \pm 6$	$48 \pm 4$	$14 \pm 3$	$9 \pm 3$
WZ+ZZ+VVV	$34 \pm 4$	$107 \pm 10$	$71 \pm 7$	$29 \pm 5$	$14 \pm 3$	$13 \pm 4$
$Z/\gamma^* \rightarrow \tau^+\tau^-$	$23 \pm 3$	$67 \pm 5$	$47 \pm 4$	$22 \pm 3$	$12 \pm 2$	$10 \pm 2$
$W\gamma^{(*)}$	$132 \pm 49$	$170 \pm 58$	$48 \pm 30$	$12 \pm 9$	$3 \pm 3$	$5 \pm 10$

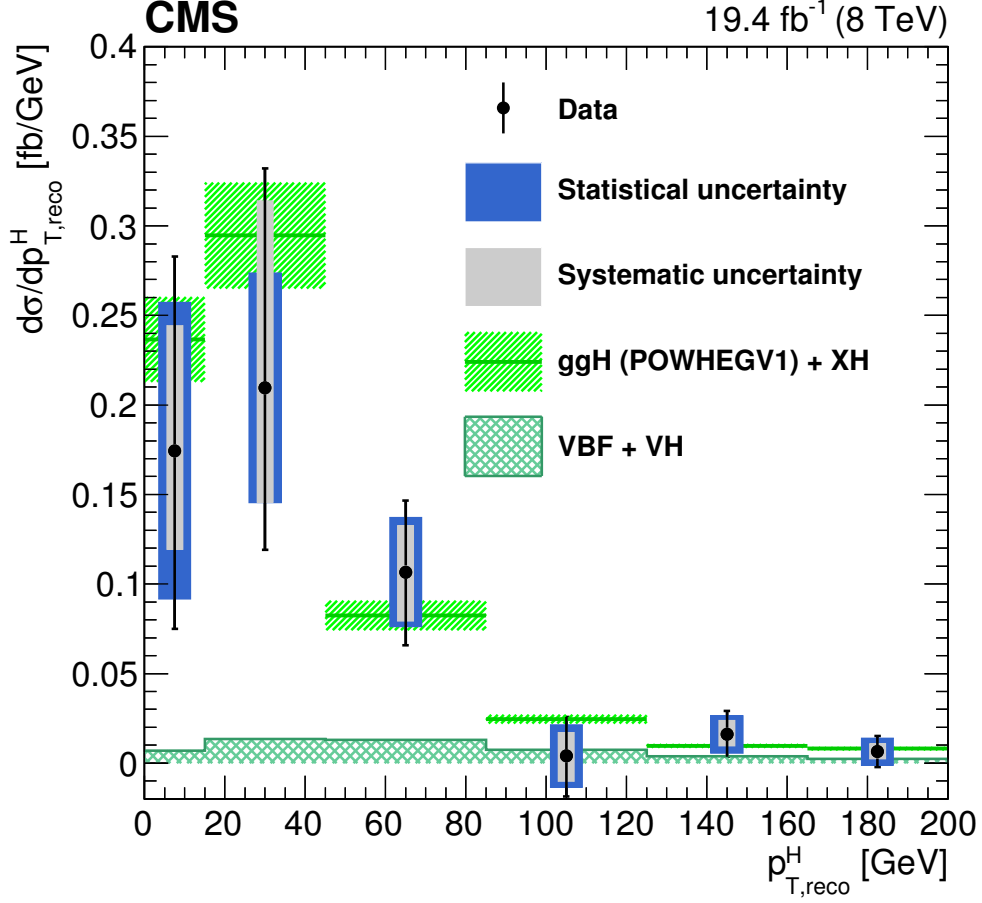
The spectrum shown in Fig. 4.23 is obtained after having performed the fit and after the subtraction of the out-of-fiducial signal events, but before undergoing the unfolding procedure. The theoretical distribution after the detector simulation and event reconstruction is also shown for comparison.

In order to assess the robustness of the fit, several toy MC samples have been produced, with a statistical accuracy corresponding to the one expected in data. The distribution of the signal strengths extracted in each bin using the toy MC samples and the their pull distributions are shown in Fig. 4.24.

## 4.7. Unfolding

To facilitate comparisons with theoretical predictions or other experimental results, the signal extracted performing the fit has to be corrected for detector resolution and efficiency effects and for the efficiency of the selection defined in the analysis. An unfolding procedure

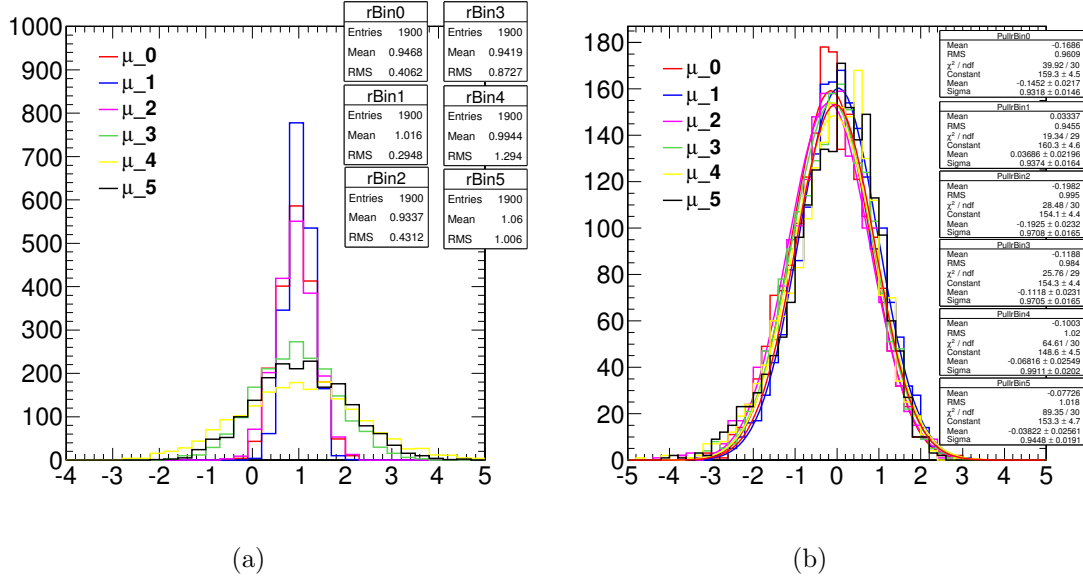




**Figure 4.23.:** Differential Higgs boson production cross section as a function of the reconstructed  $p_{T,\text{reco}}^H$ , before applying the unfolding procedure. Data values after the background subtraction are shown together with the statistical and the systematic uncertainties, determined propagating the sources of uncertainty through the fit procedure. The line and dashed area represent the SM theoretical estimates in which the acceptance of the dominant ggH contribution is modelled by POWHEG V1. The sub-dominant component of the signal is denoted as XH=VBF+VH, and is shown with the cross filled area separately.

is used relying on the ROOUNFOLD package [45], which provides the tools to run various unfolding algorithms.

The basic principle behind the unfolding procedure in this analysis is to use MC signal samples to make the “true” distribution of the variable of interest, which is obtained using simulated events before particle interaction with the detector, and the same distribution obtained using events reconstructed after the full GEANT4 simulation of the CMS detector



**Figure 4.24.:** Signal strength distribution as extracted from the fit of toy MC samples (a). Distribution of the pull of the signal strength parameters (b).

and event reconstruction. These two distributions are used to calculate the detector response matrix  $M$ :

$$R_i^{\text{MC}} = \sum_{j=1}^n M_{ij} T_j^{\text{MC}} \quad , \quad (4.23)$$

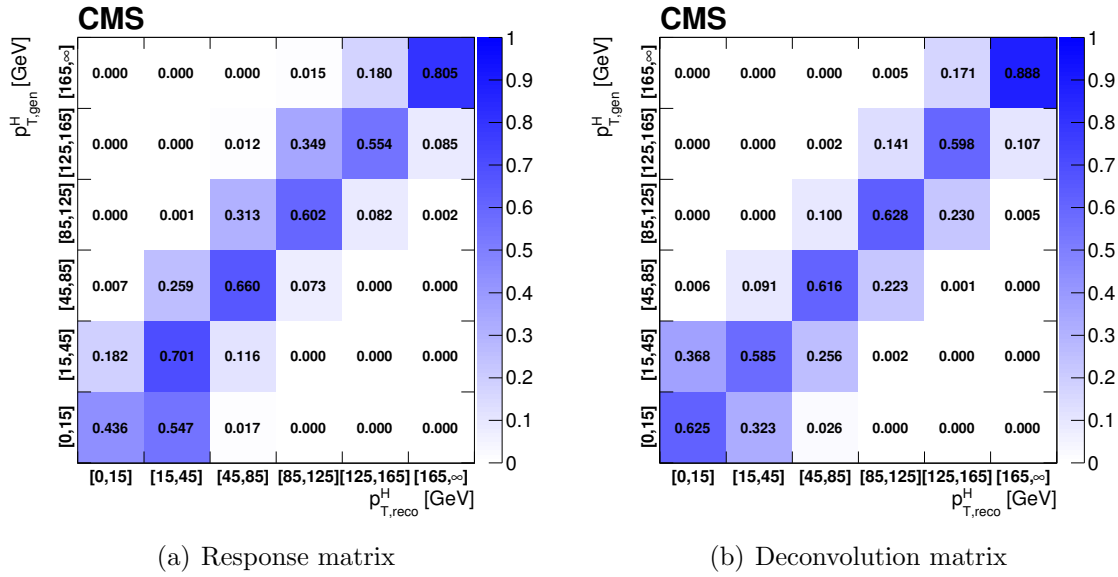
where  $R^{\text{MC}}$  and  $T^{\text{MC}}$  are two  $n$ -dimensional vectors representing the distribution before and after event processing through CMS simulation and reconstruction. The dimension  $n$  of the two vectors corresponds to the number of bins in the distributions, equal to six in this analysis. The response matrix  $M$  includes all the effects related to the detector and analysis selection that affect the  $R^{\text{MC}}$  distribution. The goal of the unfolding procedure is to obtain the  $T^{\text{truth}}$  distribution starting from the measured  $R^{\text{observed}}$  distribution by inverting the matrix  $M$ . To avoid the large variance and strong negative correlation between the neighbouring bins [19], the unfolding procedure in this analysis relies on the singular value decomposition [46] method based on the Tikhonov regularization function. Since the response matrix is in general limited by the statistical uncertainties of simulated samples and given the finite data statistical accuracy, a simple inversion could lead to

large fluctuations between bins in the unfolded result. In particular, if the off-diagonal elements of the response matrix are sizeable, the unfolded distribution has large variance and strong negative correlations between the neighbouring bins [19]. Several unfolding methods with regularization are available in literature, such as a method based on the Bayes' theorem, which overcome the unfolding instability using an iterative procedure [47]. One possible solution is the utilization of regularization methods. Such methods introduce a regularization function that controls the smoothness of the distribution and depends generally on one regularization parameter, which can be controlled to achieve the desired degree of smoothness. The choice of the regularization parameter is particularly critical, and it should represent an optimal trade-off between taming the fluctuations in the unfolded result, and biasing the unfolded distribution towards the one used to build the response matrix. The main feature of this method is the use of the singular value decomposition of the response matrix, including an additional term to suppress the oscillatory component of the solution, i.e. the regularization term, which represents some *a priori* knowledge of the final solution. The regularization parameter is chosen to obtain results that are robust against numerical instabilities and statistical fluctuations, following the prescription described in Ref. [46].

The response matrix is built as a two-dimensional histogram, with the generator-level  $p_T^H$  on the  $y$  axis and the same variable after the reconstruction on the  $x$  axis, using the same binning for both distributions. The resulting detector response matrix, including all signal sources and normalized by row, is shown in Fig. 4.25(a). The value of the diagonal bins corresponds to the stability  $S$ . The same matrix, normalized by column, is shown in Fig. 4.25(b). In this case the diagonal bins correspond to the purity  $P$ . The  $S$  and  $P$  parameters, defined in Sec. 4.3, provide an estimate of the  $p_T^H$  resolution and migration effects. The main source of bin migrations effects in the response matrix is the limited resolution in the measurement of  $E_T^{\text{miss}}$ .

The resulting detector response matrix, which includes the effects of all signal sources and is represented by normalizing each row to unity is shown in Fig. 4.25(a). This representation shows the stability  $S$  in the diagonal bins, where  $S$  is defined as the ratio of the number of events generated and reconstructed in a given bin, and the number of events generated in that bin. In addition, a deconvolution matrix is constructed by normalizing each column to unity and is shown in Fig. 4.25(b). This latter representation shows the purity  $P$  in the diagonal bins, where  $P$  is defined as the ratio of the number of events generated and

reconstructed in a given bin, and the number of events reconstructed in that bin. The  $S$  and  $P$  parameters provide an estimate of the  $p_T^H$  resolution and of migration effects. The response matrix built including all signal sources is shown in Fig. 4.25. In order to point out either the purity or the stability in diagonal bins, each column or row of the matrix was respectively normalized to unity. The matrix obtained in the first case is what is actually called detector response matrix, while in the other case the matrix is usually referred to as detector deconvolution matrix.



**Figure 4.25.:** Response matrix (a) and deconvolution matrix (b) including all signal processes. The matrices are normalized either by row (a) or by column (b) in order to show the purity or stability respectively in diagonal bins.

Several closure tests are performed in order to validate the unfolding procedure. To estimate the uncertainty in the unfolding procedure due to the particular model adopted for building the response matrix, two independent gluon fusion samples are used, corresponding to two different generators: POWHEG V1 and JHUGEN generators, both interfaced to PYTHIA 6.4. The JHUGEN generator sample is used to build the response matrix while the POWHEG V1 sample is used for the measured and the MC distributions at generator level. The result of this test shows good agreement between the unfolded and the distribution from MC simulation.

In order to further prove the choice of the regularization parameter, a large number of simulated pseudo-experiments has been generated to verify that the coverage of the unfolded

**Table 4.8.:** Coverage interval for each bin and for different values of the regularization parameter, obtained using pseudo-experiments.

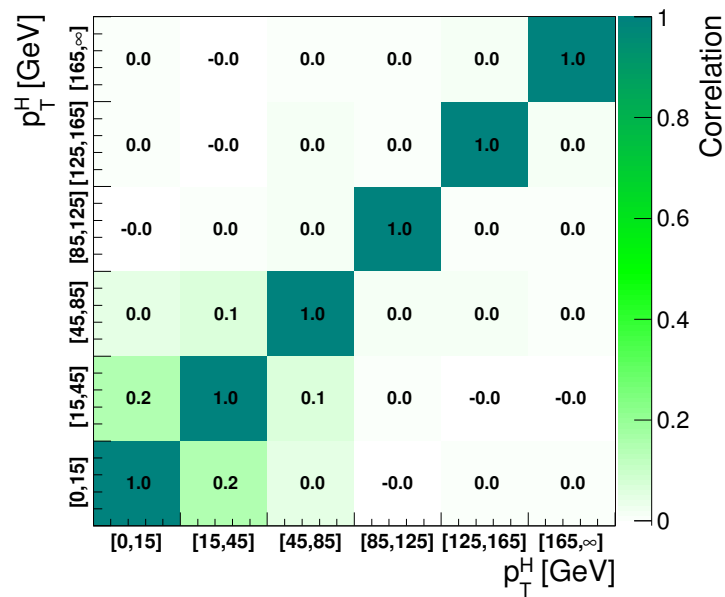
$p_T^H$ bin [GeV]	Coverage			
	$k_{\text{reg}} = 2$	$k_{\text{reg}} = 3$	$k_{\text{reg}} = 4$	$k_{\text{reg}} = 5$
0–15	$0.654^{+0.015}_{-0.016}$	$0.704^{+0.015}_{-0.015}$	$0.727^{+0.014}_{-0.015}$	$0.755^{+0.014}_{-0.014}$
15–45	$0.701^{+0.015}_{-0.015}$	$0.665^{+0.015}_{-0.016}$	$0.683^{+0.015}_{-0.015}$	$0.733^{+0.014}_{-0.015}$
45–85	$0.717^{+0.014}_{-0.015}$	$0.706^{+0.015}_{-0.015}$	$0.709^{+0.015}_{-0.015}$	$0.716^{+0.014}_{-0.015}$
85–125	$0.634^{+0.016}_{-0.016}$	$0.681^{+0.015}_{-0.015}$	$0.714^{+0.015}_{-0.015}$	$0.739^{+0.014}_{-0.015}$
125–165	$0.599^{+0.015}_{-0.016}$	$0.650^{+0.015}_{-0.016}$	$0.700^{+0.015}_{-0.015}$	$0.751^{+0.014}_{-0.014}$
165– $\infty$	$0.632^{+0.016}_{-0.016}$	$0.674^{+0.015}_{-0.015}$	$0.701^{+0.015}_{-0.015}$	$0.722^{+0.014}_{-0.015}$

uncertainties obtained with this procedure is as expected. From each pseudo-experiment the reconstructed  $p_T^H$  spectrum is obtained and then unfolded using the procedure described above, including only the statistical uncertainties. The coverage is calculated for each  $p_T^H$  bin, counting the number of pseudo-experiments for which the statistical uncertainty covers the true value. The confidence intervals are calculated using the Clopper-Pearson approach, and the results are shown in Table 4.8 for different values of the regularization parameter: starting from  $k_{\text{reg}} = 2$  (stronger regularization) up to  $k_{\text{reg}} = 5$  (weaker regularization). The criterion for choosing the best  $k_{\text{reg}}$  value is to increase the regularization as much as possible without introducing a bias, i.e. until a 68% coverage is fulfilled. This criterion leads to the same result as the prescription described in Ref. [46], strengthening the choice of  $k_{\text{reg}} = 3$ .

#### 4.7.1. Treatment of systematic uncertainties

An important aspect of this analysis is the treatment of the systematic uncertainties and the error propagation through the unfolding procedure. The sources of uncertainty are divided into three categories, depending on whether the uncertainty affects only the signal yield (type A), both the signal yield and the response matrix (type B), or only the response matrix (type C). These three classes propagate differently through the unfolding procedure.

Type A uncertainties are extracted directly from the fit in the form of a covariance matrix, which is passed to the unfolding tool as the covariance matrix of the measured distribution. The nuisance parameters belonging to this category are the background shape



- the b veto scale factor. It affects the signal and background templates by varying the number of events with jets that enter the selection. It also affects the response matrix because the reconstructed spectrum is harder or softer depending on the number of jets, which in turn depends on the veto.
- the lepton efficiency scale factor. It affects the signal and background template shape and normalization. It affects the response matrix by varying the reconstructed spectrum;
- the  $E_{\text{T}}^{\text{miss}}$  scale and resolution, which have an effect similar to the above;
- lepton scale and resolution. The effect is similar to the above;
- jet energy scale. It affects the signal and background template shape and normalization. It also affects the response matrix because, by varying the fraction of events with jets,

the b veto can reject more or fewer events, thus making the reconstructed spectrum harder or softer.

The effect of each type B uncertainty is evaluated separately, since each one changes the response matrix in a different way. In order to evaluate their effect on the signal strengths parameters, two additional fits are performed, each time fixing the nuisance parameter value to  $\pm 1$  standard deviation with respect to its nominal value. The results of the fits are then compared to the results of the full fit obtained by floating all the nuisance parameters, thus determining the relative uncertainty on the signal strengths due to each nuisance parameter, as shown in Tab. 4.9. Using these uncertainties, the measured spectra for each type B source are built. The effects are propagated through the unfolding by building the corresponding variations of the response matrix and unfolding the measured spectra with the appropriate matrix.

Type C uncertainties are related to the underlying assumption on the Higgs boson production mechanism used to extract the fiducial cross sections. These are evaluated using alternative response matrices that are obtained by varying the relative fraction of the VBF and ggH components within the experimental uncertainty, as given by the CMS combined measurement [48]. Three different response matrices are built, corresponding to the nominal, scaled up, and scaled down VBF/ggH ratio. The nominal matrix assumes the SM VBF/ggH ratio, while up- and down-scaled matrices are constructed by varying the SM signal strengths within the experimental constraints for VBF and ggH in such a way as to obtain the maximal variation of the VBF/ggH ratio allowed by the experimental constraints. These three matrices are used to unfold the reconstructed spectrum with the nominal VBF/ggH fraction, and obtain an uncertainty on the unfolded spectrum.

## 4.8. Results

Once the signal strengths are extracted from the fit results, as explained in section ??, and the uncertainties are decoupled the categories depicted in section ??, i.e. type A and type B uncertainties, we can go on unfolding the measured spectrum. In figure 4.27 is shown the  $p_{T,\text{reco}}^H$  differential distribution before applying the unfolding procedure, compared with the MC truth expectation. The corresponding numbers are reported in table 4.10.

**Table 4.9.:** Effect of all the Type B uncertainties on the signal strengths of each bin. In the table are reported the signal strength variations corresponding to an up or down scaling of each nuisance.

Type B uncertainty	Effect on signal strength ( $+1\sigma/ -1\sigma$ [%])					
	[0–15]	[15–45]	[45–85]	[85–125]	[125–165]	[165– $\infty$ ]
b veto	-10.1/-8.8	7.3/12.2	-6.3/3.1	-14.4/-4.8	-5.4/14.5	-7.9/17.8
lepton efficiency	-14.7/-3.9	4.5/15.1	-5.7/2.5	-13.2/-5.3	-0.2/7.6	-0.1/6.8
$E_T^{\text{miss}}$ resolution	-12.5/0.0	15.4/-0.0	-12.8/-0.0	8.7/0.0	-20.9/-0.0	10.5/0.0
$E_T^{\text{miss}}$ scale	-14.4/-6.8	-0.0/17.7	-6.1/-7.1	9.6/-20.9	2.3/32.4	2.5/2.6
lepton resolution	-12.5/-0.0	11.2/0.0	-2.4/0.0	-13.4/-0.0	9.9/0.0	-4.6/-0.0
electron momentum scale	-2.7/-13.1	15.9/9.9	10.8/-16.8	16.2/-33.1	30.9/-14.4	12.6/-10.9
muon momentum scale	-7.0/-10.7	11.8/8.9	1.1/-8.7	-0.7/-14.4	14.5/-4.6	8.0/-1.6
jet energy scale	-10.9/-10.1	9.0/9.0	-3.0/-2.9	-10.3/-8.9	0.3/3.4	5.2/3.1

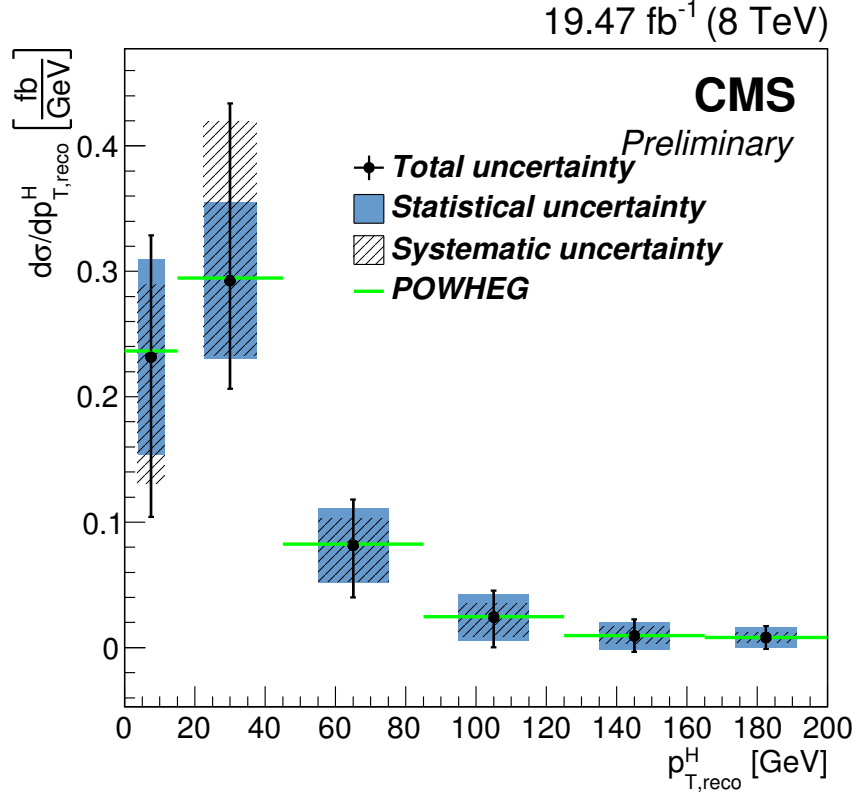
**Table 4.10.:** Measured values for each bin of  $p_T^H$  with the corresponding total uncertainty compared with the MC truth expectation.

Bin	Unfolded value	Total error(%)	Stat error(%)	Syst error(%)	MC truth
1	0.23	+41.9/-55.0	$\pm 33.6$	+25.0/-43.5	0.24
2	0.30	+48.3/-29.5	$\pm 21.3$	+43.3/-20.4	0.30
3	0.08	+44.8/-50.9	$\pm 36.5$	+26.1/-35.4	0.08
4	0.02	+88.1/-99.1	$\pm 75.9$	+44.8/-63.8	0.02
5	0.01	+141.1/-135.2	$\pm 116.3$	+80.0/-68.9	0.01
6	0.01	+112.9/-110.6	$\pm 99.5$	+53.3/-48.4	0.01

In order to unfold the spectrum, the procedure described in section 4.7 has been pursued. The statistical plus type A systematic uncertainties are correctly propagated by the unfolding procedure into the final spectrum, taking into account the signal strengths covariance matrix. The type B systematic uncertainty has been propagated using the following procedure: for each  $p_T^H$  bin, we compute the upper bound of the systematic band computing the square sum of all the signal strength variations that deviate in the up direction with respect to the bin central value, whether or not this variation correspond to the up or down shift of the nuisance. The same is done for the lower bound of the systematic band. If both the up and down shifts of a given nuisance leads to a same direction variation of the signal strength, only the larger variation is considered.

Results are reported in terms of a differential distribution, dividing by the luminosity (the





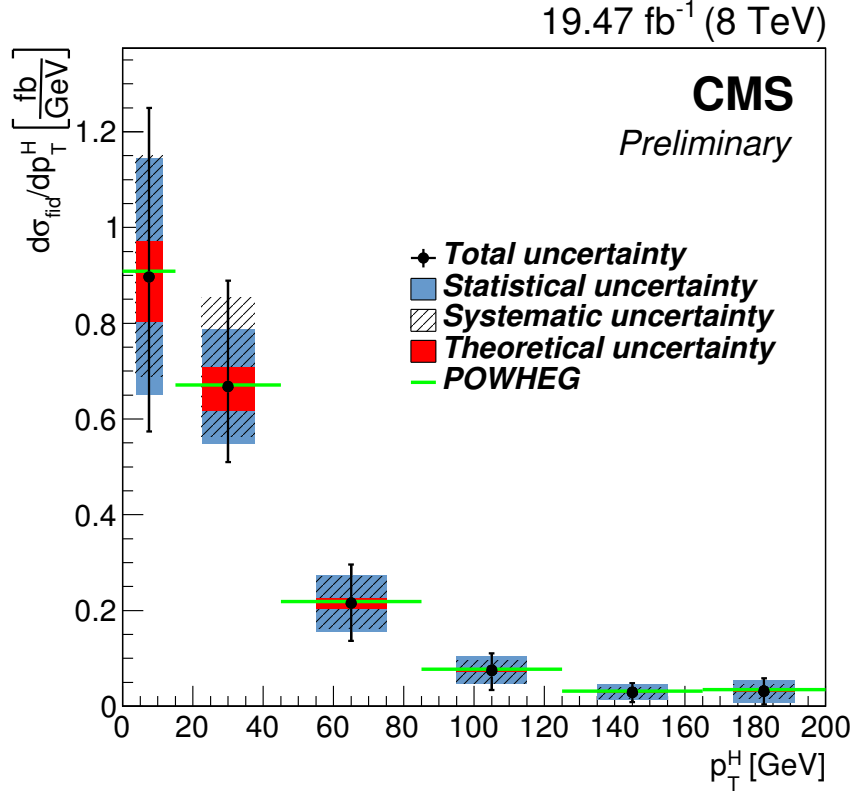
**Figure 4.27.:** Differential Higgs production cross section as a function of  $p_{T, \text{reco}}^H$  before applying the unfolding procedure. The bins content corresponds to the MC prediction since the fit is performed on an Asimov dataset. The MC truth is represented by the green line.

uncertainty on the luminosity measurement is included in the statistical error band), and putting in each bin the bin content divided by the bin width.

In figure 4.28 is shown the Higgs differential production cross section as a function of  $p_T^H$ . The results reported in this plot have been extracted fitting an Asimov dataset, thus the bins content corresponds to what expected from the MC predictions. The red shaded area in each bin represents the total uncertainty due to statistics plus systematics (type A and B). The blue area corresponds to the statistical error only. Data are not shown since the analysis is still blinded. As a cross check, the MC truth prediction has been superimposed to the plot. The final results are also shown in table 4.11. The systematic error reported in the table has been extracted in each bin taking the difference of the squares of total and statistical error.

A comparison between pre-unfolding and unfolded distributions shows that the relative

uncertainties in each bin get reduced after the unfolding. To check the correctness of this result, a test using MC toys has been performed and is explained in Appendix ??.



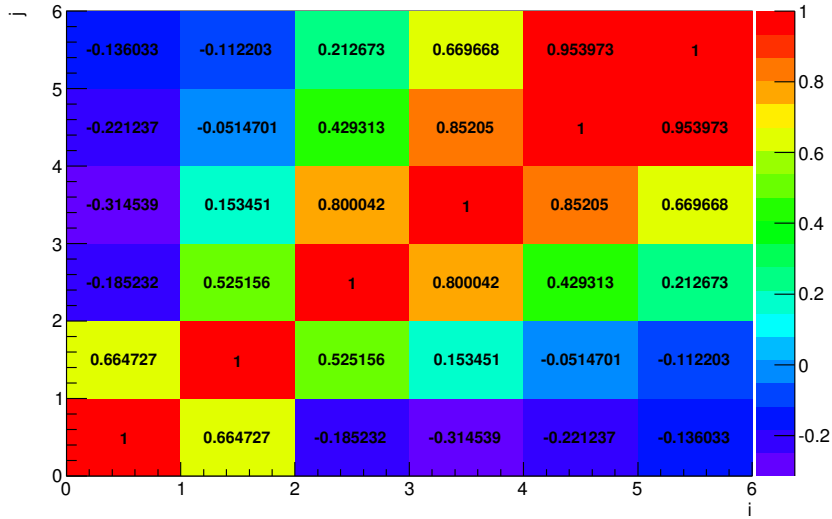
**Figure 4.28.:** Unfolded differential Higgs production cross section as a function of  $p_T^H$ . The bins content corresponds to the MC prediction since the fit is performed on an Asimov dataset. The error bars are the total expected uncertainties in this measurement. Also the statistical, the systematic and the theoretical uncertainties are shown separately. The MC truth (POWHEG) is represented by the green histogram.

Together with the unfolded spectrum, also the covariance matrix of the six bins is reported, in order to assess how much different bins are correlated. The result is shown in figure 4.29.

The differential spectrum can be integrated to obtain a measurement of the inclusive cross section inside the fiducial region. The uncertainties can be correctly taken into account in this calculation using the covariance matrix of the six signal strengths to propagate the errors. In this case the unfolding procedure is not needed and to extrapolate the measured

**Table 4.11.:** Unfolded values for each bin of  $p_T^H$  with the corresponding total uncertainty compared with the MC truth expectation.

Bin	Unfolded value	Total error(%)	Stat error(%)	Syst error(%)	MC truth
1	0.23	+42.4/-55.4	$\pm 34.0$	+25.3/-43.7	0.24
2	0.29	+48.7/-30.0	$\pm 21.5$	+43.6/-20.9	0.29
3	0.08	+45.7/-51.9	$\pm 36.9$	+26.9/-36.4	0.08
4	0.02	+88.5/-99.5	$\pm 76.1$	+45.2/-64.1	0.02
5	0.01	+141.3/-135.3	$\pm 116.4$	+80.1/-69.0	0.01
6	0.01	+112.9/-110.6	$\pm 99.5$	+53.4/-48.4	0.01



**Figure 4.29.:** Covariance matrix of the six  $p_T^H$  bins related to the unfolded distribution.

result to the fiducial region, only the efficiency of the analysis selection is needed.

The measured cross section after the analysis selection, i.e. number of events divided by luminosity, is  $14 \pm 4$  fb. Using the overall efficiency defined in section ??, i.e.  $\epsilon = 0.362 \pm 0.005$ , the cross section value in the fiducial region can be determined and is equal to:

$$\sigma_{fid} = 47 \pm 7 \text{ (stat)} \pm 9 \text{ (syst)} \text{ fb} \quad . \quad (4.24)$$

As a closure test, the measurement can be extrapolated to the full  $4\pi$  acceptance, using the efficiency reported in section ??, which is  $\epsilon = 0.03960 \pm 0.00033$ . The resulting cross section is:

$$\sigma_{4\pi} = 433 \pm 67 \text{ (stat)} \pm 83 \text{ (syst)} \text{ fb} \quad , \quad (4.25)$$

in agreement with the expected value from MC.

## Chapter 5.

### First $H \rightarrow WW$ results at 13 TeV

#### 5.1. Higgs boson search at 13 TeV

#### 5.2. Search for a high mass resonance in the $WW$ decay channel at 13 TeV



## Chapter 6.

## Conclusions





# Appendix A.

## Fiducial region definition and optimization

The fiducial region must be chosen in such a way to be as close as possible to the selections applied in the analysis, in order to reduce the model dependence in the extrapolation step. That means that for optimizing the fiducial volume definition, the efficiency has to be maximized. Another parameter entering the game is the number of fake events, in other words the number of reconstructed events which do not belong to the fiducial phase space. This parameter should instead be as small as possible. Even if we have to observe the trend of these two quantities as a function of  $p_T^H$ , we can maximize the ratio between the overall efficiency and the overall fake rate as a proxy for establishing the “goodness” of the fiducial region.

Several different fiducial region definitions were tested and the results show that:

- **of cut:** The fiducial region definition must include only the opposite flavor combination including one electron and one muon. If we include also the combinations involving  $\tau$ 's the efficiency falls down.
- **Lepton cut:** Since the resolution on lepton transverse momentum is good, there is no need to loosen the cuts related these variables, i.e. we can use the same cuts defined in the analysis selection ( $p_T^{\ell,1} > 20 \text{ GeV}$ ,  $p_T^{\ell,2} > 10 \text{ GeV}$ ).
- **Di-lepton  $p_T$  cut:** As stated in the previous point, there is no need to loosen this cut, so we kept the same value as the analysis selection, i.e.  $p_T^{\ell\ell} > 30 \text{ GeV}$ .

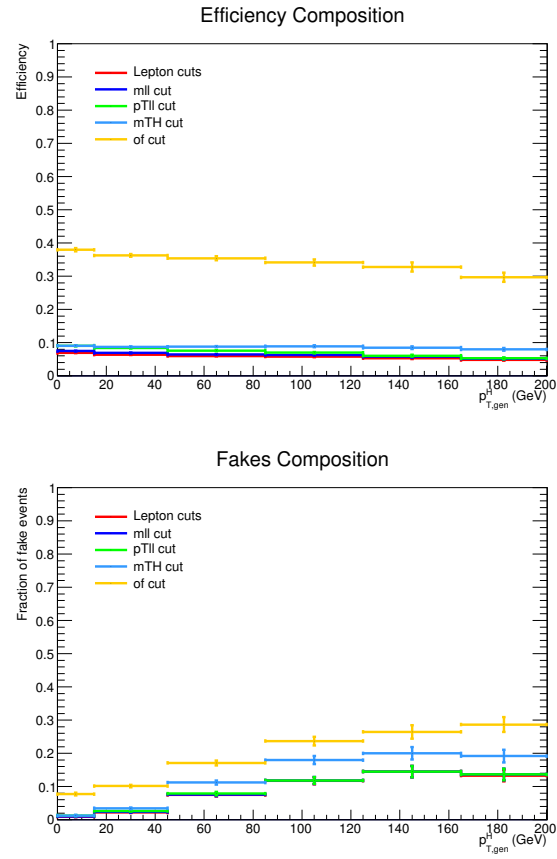
- **Di-lepton mass cut:**  $m_{\ell\ell} > 12 \text{ GeV}$  as discussed before.
- **neutrino pair  $p_T$  cut:** Since the resolution on the measurement of the missing transverse energy is poor, the neutrino pair cut should not be included in the definition of the fiducial region, because it would increase the fake rate without increasing the efficiency, thus resulting in a lower ratio between overall efficiency and fake rate.
- **$m_T$  cut:** Also the  $m_T$  cut that we have in the analysis selection, i.e.  $m_T > 60 \text{ GeV}$ , should be loosened or removed because it involves neutrinos and then increase the fake rate. We decided eventually to keep this cut, loosening it to  $50 \text{ GeV}$ , because in addition to increase the number of fake events, it increases the efficiency as well.

The fake rate and the efficiency as a function of  $p_T^H$  after the optimization discussed before are shown in figure A.1. To obtain these plots the fiducial region was modified adding in sequence the various cuts and computing the efficiency and the fake rate each time. In that way we can assess the composition of those distributions.

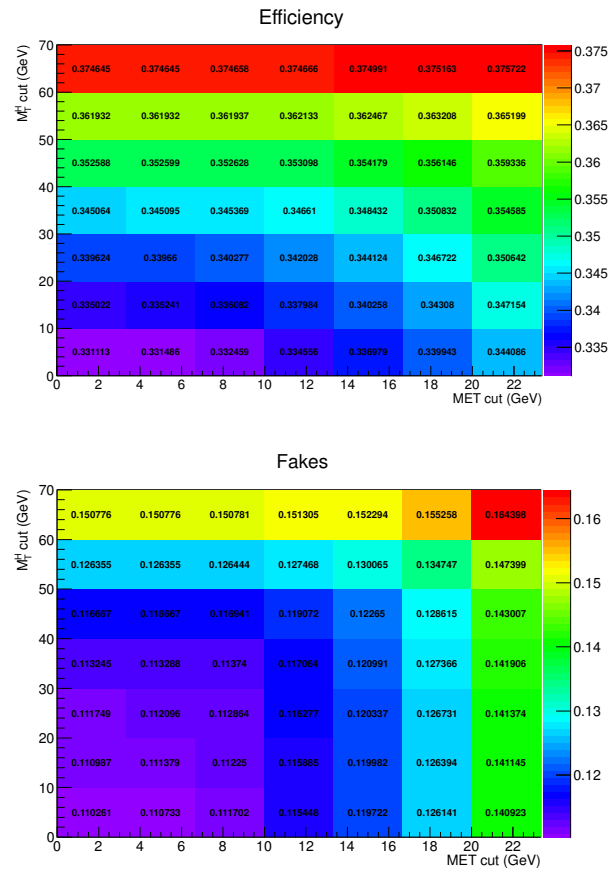
The efficiency and fraction of fake events have been measured also as a function of the  $E_T^{\text{miss}}$  and  $m_T$  cuts in the fiducial region. Since these two variables are correlated, the results are reported as two-dimensional histograms. In Fig. A.2 are reported the efficiency and fraction of fake events for these two variables.

The criterion adopted to define the fiducial region is a tradeoff between having a large efficiency and a small fraction of fake events. Especially when looking at the low resolution variables, such as  $E_T^{\text{miss}}$  and  $m_T$ , a suitable figure of merit has to be chosen for the estimation of the best cuts. Several different figures of merit have been checked, such as  $\epsilon/f$ ,  $\epsilon - f$  and  $(1 - f)/\epsilon$ . The results for these three different figures of merit are shown in Fig. A.3 as a function of the  $E_T^{\text{miss}}$  and  $m_T$  cuts in the fiducial region.

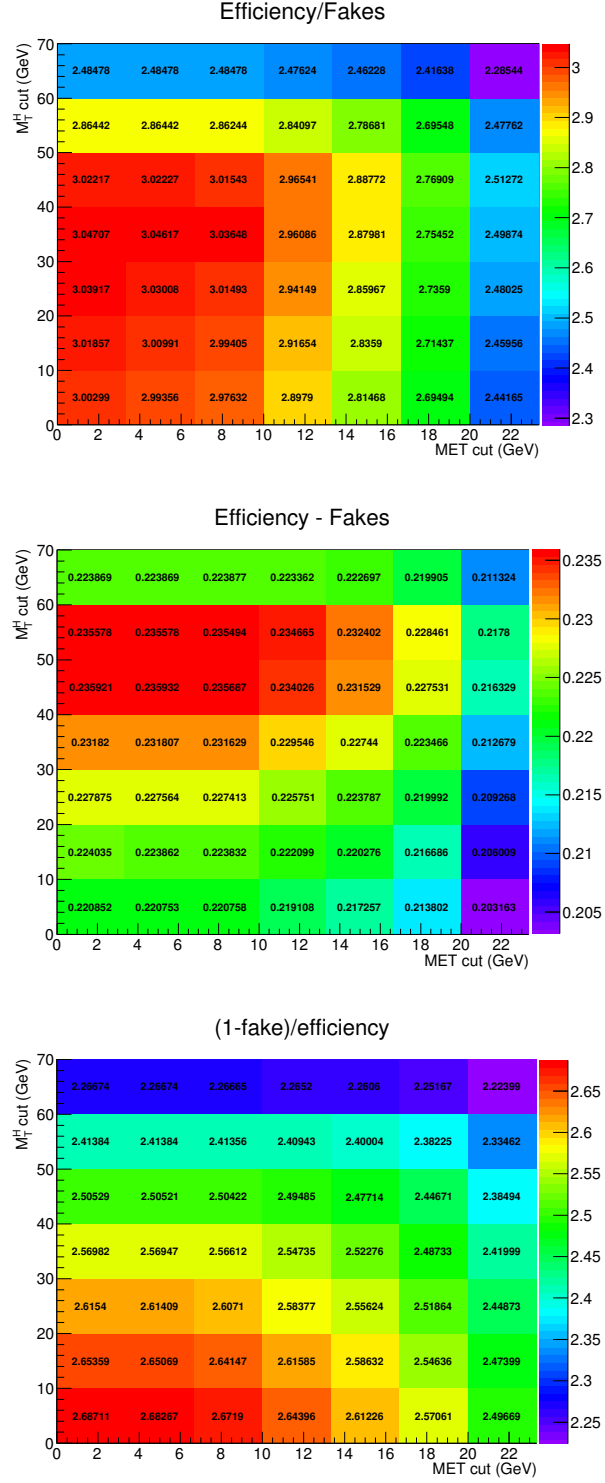
Following the same criterion, similar plots as above have been obtained for an alternative model, given by varying up the ggH/VBF ratio within the experimental uncertainties. The results, shown in Fig. A.4 and Fig. A.5, show a similar trend with respect to the model with nominal ggH/VBF ratio.



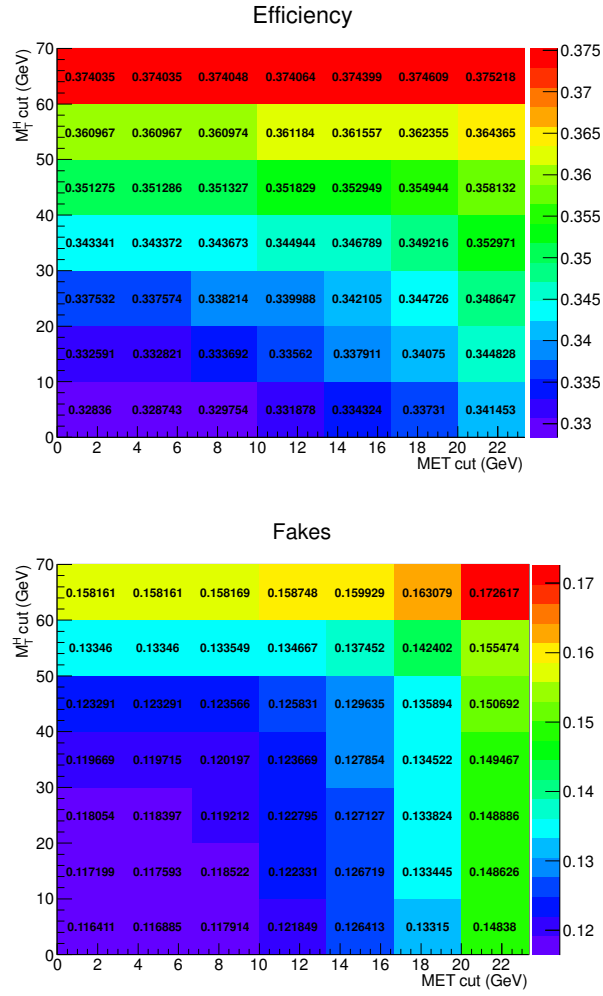
**Figure A.1.:** Efficiency and fake rate as a function on Higgs transverse momentum. The plots correspond to the optimized fiducial region definition and show the effect of adding each of the mentioned cuts in sequence.



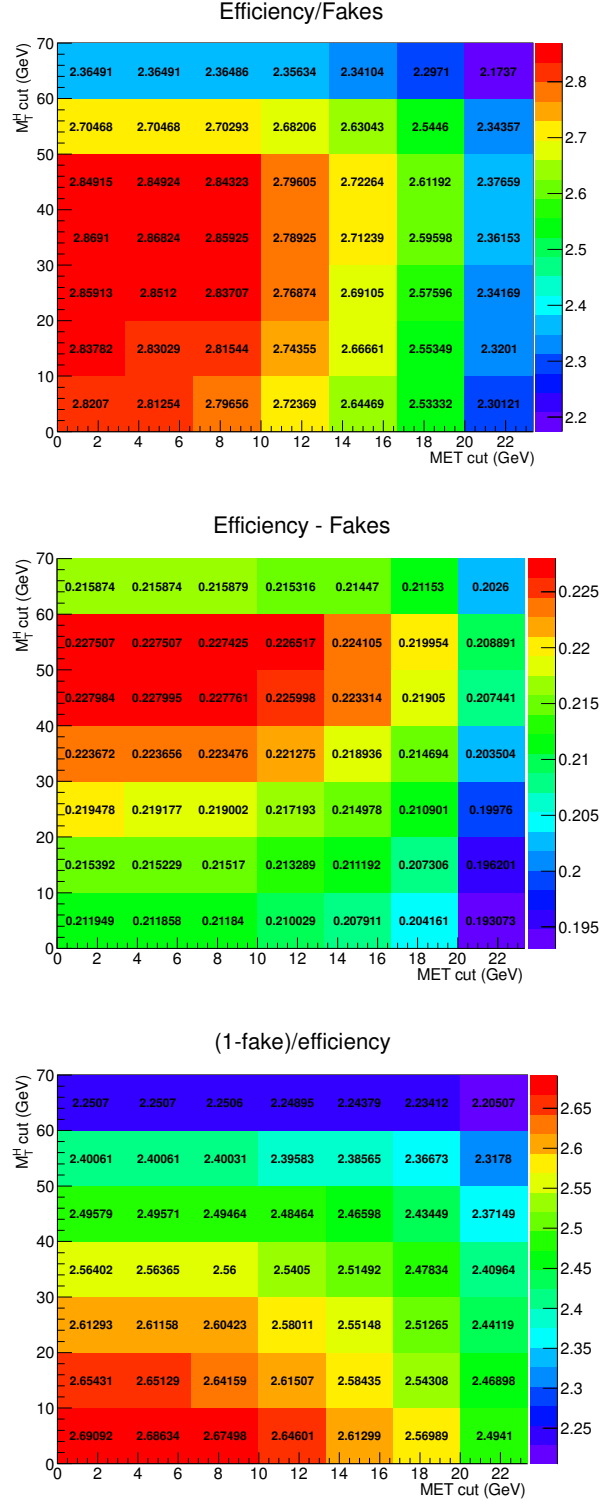
**Figure A.2.:** Efficiency and fake rate as a function of  $E_T^{\text{miss}}$  and  $m_T$  cuts in the fiducial region.



**Figure A.3.:** Different figures of merit as a function of  $E_T^{\text{miss}}$  and  $m_T$  cuts in the fiducial region.



**Figure A.4.:** Efficiency and fake rate as a function of  $E_T^{\text{miss}}$  and  $m_T$  cuts in the fiducial region, for the alternative model with an up variation of the ggH/VBF ratio.



**Figure A.5.:** Different figures of merit as a function of  $E_T^{\text{miss}}$  and  $m_T$  cuts in the fiducial region, for the alternative model with an up variation of the ggH/VBF ratio.









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