STANDARD MODEL IS BEST MODEL (WORKING TITLE)

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- 28 template.

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ABSTRACT STANDARD MODEL IS BEST MODEL (WORKING TITLE) William Kennedy DiClemente J. Kroll

This is the abstract text.

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109 This is the preface. It's optional, but it's nice to give some context for the reader and stuff.

Will K. DiClemente Philadelphia, February 2019

Introduction

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The Standard Model (SM)¹ has been remarkably successful...

¹Here's a footnote.

Theoretical Framework

(Some example introductory text for this chapter)...

117 2.1 Introduction to the Standard Model

- Modern particle physics is generally interpreted in terms of the Standard Model (SM). This is a
- quantum field theory which encapsulates our understanding of the electromagnetic, weak, and strong
- 120 interactions...

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2.2 Electroweak Mixing and the Higgs Field

- When the theory of the electroweak interaction was first developed [1, 2], the W and Z bosons were
- predicted to be massless (a typical mass term in the Lagrangian would violate the SU(2) symmetry).
- However, these were experimentally observed to have masses...

LHC and the ATLAS Detector

127 3.1 The Large Hadron Collider

128 The Large Hadron Collider (LHC) [3] is...

129 3.2 The ATLAS Detector

125

126

130 ATLAS is a general-purpose particle detector...

3.2.1 The Inner Detector

132 The Inner Detector serves the primary purpose of measuring the trajectories of charged particles...

3.2.1.1 Pixel Detector

The Pixel detector consists of four cylindrical barrel layers and three disk-shaped endcap layers...

135 3.2.1.2 Semiconductor Tracker

- 136 The Semiconductor Tracker uses the same basic technology as the Pixels, but the fundamental unit
- of silicon is a larger "strip"...

3.2.1.3 Transition Radiation Tracker

The Transition Radiation Tracker is the outermost component of the ID...

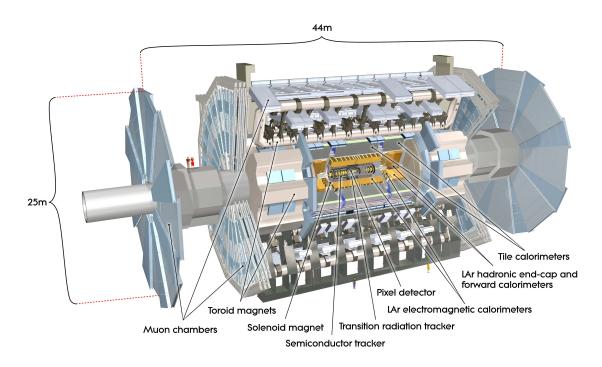


Figure 3.1: General cut-away view of the ATLAS detector [4].

140 3.2.2 The Calorimeters

- 141 ATLAS includes two types of calorimeter system for measuring electromagnetic and hadronic show-
- ers. These are the Liquid Argon (LAr) calorimeters and the Tile calorimeters. Together, these cover
- the region with $|\eta| < 4.9...$

3.2.2.1 Liquid Argon Calorimeters

145 The Liquid Argon system consists of...

146 3.2.2.2 Tile Calorimeters

147 The Tile calorimeter provides coverage for hadronic showers...

Alignment of the ATLAS Inner Detector

In order for the subdetectors of the ID to operate at their designed precisions, it is essential that the locations of the sensors be known as precisely as possible. Differences between the expected and actual positions of a sensor can result in displaced particle hits and degrade track reconstruction quality. These misalignments can occur for any number of reasons, including but not limited to elemnts shifting during maintenance periods or cycles in ATLAS's magnetic field, or simply small movements during normal detector operations. Since it is not practical to physically realign hundreds of thousands of detector elements to μ m precision by hand, an iterative track-based alignment algorithm is used to determine the physical positions and orientations of these elements [5]. The effects of misalignments and the steps taken to correct and monitor them are detailed in this chapter.

4.1 Effects of Misalignment

160 Hello world!

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61 4.2 The Alignment Method

162 Hello world!

4.3 Momentum Bias Corrections

165 4.4 Alignment of the IBL

166 Hello world!

167 4.5 Alignment Monitoring

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WZ production @ $\sqrt{s} = 13$ TeV

171 5.1 Theoretical motivation

172 Hello world!

5.2 Signal definition

174 Hello world!

175 5.3 Background estimations

176 Hello world!

5.4 Cross section measurement

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Same-sign $WW @ \sqrt{s} = 13 \text{ TeV}$

181 6.1 Theoretical motivation

182 Hello world!

183 6.2 Signal definition

184 Hello world!

185 6.3 Background estimations

186 Hello world!

187 6.4 Cross section measurement

Prospects for same-sign WW at the High Luminosity LHC

On December 3, 2018, Run 2 of the LHC officially ended, and the collider was shut down to begin the first of two scheduled extended maintenance periods [6]. During these two long shutdowns, the Phase-I and Phase-II upgrades of the LHC and ATLAS will occur in order to prepare for the High-Luminosity LHC (HL-LHC) which is scheduled to begin operation in 2026 [7]. The HL-LHC is planned to run at an instantaneous luminosity of $\mathcal{L} = 5 \times 10^{34}$ cm⁻²s⁻¹ with an

The HL-LHC is planned to run at an instantaneous luminosity of $\mathcal{L} = 5 \times 10^{54}$ cm⁻²s⁻¹ with an average of 140 collisions per beam-crossing. Over the course of operation, the HL-LHC is expected to collect a total integrated luminosity of $\mathcal{L} = 3000$ fb⁻¹ by 2035 [8].

These run conditions are much harsher than what ATLAS has experienced so far, and as a result there are several planned upgrades to the detector. Most notably, the entire ID will be replaced with an all-silicon tracker which will extend the coverage from $|\eta| \leq 2.7$ up to $|\eta| \leq 4.0$. This will allow for reconstruction of charged particle tracks which can in turn be matched to clusters in the calorimeters for electron identification or forward jet tagging [9].

TODO: Why are we studying ssww at the HL-LHC

7.1 Theoretical motivation

The theoretical motivation for studying the ssWW process is detailed in Section 6.1.

7 7.2 Signal definition

Hello world!

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9 7.2.1 Sensitivity to longitudinal polarization

210 7.3 Background estimations

211 Hello world!

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212 7.4 Selection optimization

As mentioned earlier, the HL-LHC will feature forward tracking, an increase in center of mass energy, and a higher integrated luminosity. Therefore, this study is an excellent time to see if there are new optimizations to the signal event selection that can improve the signal to background ratio.

216 7.4.1 Random grid search algorithm

The chosen algorithm for optimizing the event selection is known as the Random Grid Search 217 (RGS) [10]. Consider a simple case of two variables x and y chosen to differentiate the signal from 218 the background. In order to be considered a signal event, a given event would be required to pass 219 a cut point $c = \{x > x_c, y > y_c\}$. A simple method to choose the optimal cut point (i.e. the 220 "best" values of the cuts x_c and y_c) would be to construct an $n \times m$ rectangular grid in x and y221 consisting of points $(x_0, y_0), (x_1, y_1), ..., (x_n, y_m)$, as in Figure 7.1. One can then choose a cut point 222 $c_k = \{x > x_i, y > y_j\}$ that maximizes the signal significance as measured by a chosen metric. This 223 would be considered a regular or rectangular grid search. 224

While effective in principle, this rectangular grid search comes with two major drawbacks:

- 1. The algorithm does not scale well as the number of variables to be optimized—the dimensionality of the grid—increases. In the case of a square grid with N bins per variable v, the number of cut points to be evaluated grows as N^v .
- Signal and background samples are rarely evenly distributed over the entire grid, resulting in many cut points being sub-optimal and evaluating them would be a waste of computing resources.

To combat these limitations, the RGS algorithm constructs a grid of cut points directly from the signal sample itself. In the two-dimensional example, this means that the variables x_i and y_j making up the cut point $c_k = \{x > x_i, y > y_j\}$ take their values directly from a given signal event. This has the benefit of creating a random grid of cut points that is by construction biased towards regions of high signal concentration. This reduces the need for exponentially increasing numbers of cut points while ensuring that computing resources are not wasted in regions with few to no signal events. An example of the two-dimensional random grid is shown in Figure 7.2.

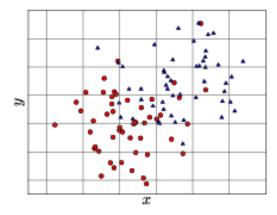


Figure 7.1: A visual representation of a rectangular grid search algorithm. The signal events are the blue triangles, and the red circles are the background events. TODO: replace with own figure

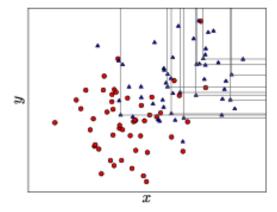


Figure 7.2: A visual representation of a random grid search algorithm. The signal events are the blue triangles, and the red circles are the background events. TODO: replace with own figure

Once the random grid of cut points is constructed, the optimal cut point can be chosen using
whatever metric the analyzer chooses, such as signal to background ratio. For the purpose of the $W^{\pm}W^{\pm}jj$ upgrade study, the optimal cut point is the one that mazimizes the signal significance Z

defined as in Equation 7.1 [11].

$$Z = \sqrt{2\left[(s+b)\ln\left(\frac{s+b}{b_0}\right) + b_0 - s - b\right] + \frac{(b-b_0)^2}{\sigma_b^2}}$$
 (7.1)

where s and b are the number of signal and background events, respectively, σ_b is the total uncertainty on the background, and b_0 is defined as:

$$b_0 = \frac{1}{2} \left(b - \sigma_b^2 + \sqrt{(b - \sigma_b^2)^2 + 4(s + b)\sigma_b^2} \right)$$
 (7.2)

In the case where the backround is known precisely (i.e. $\sigma_b = 0$), Equation 7.1 simplifies to

$$Z = \sqrt{2\left(b\left[(1+s/b)\ln(1+s/b) - s/b\right]\right)}$$
 (7.3)

which further reduces to the familiar $Z = s/\sqrt{b}$ for the case when s << b.

7.4.2 Inputs to the optimization

- The RGS algorithm was trained on the sample of LL-polarized $W^{\pm}W^{\pm}jj$ events, and the TX-
- 249 polarized sample was treated as a background for the purpose of the optimization even though it
- is a subset of the $W^{\pm}W^{\pm}jj$ electroweak signal in order to enhance the longitunal polarization as
- 251 much as possible.
- The random grid was constructed from nearly 40,000 LL signal events and
- The variables chosen for optimization are:
- Leading lepton $p_{\rm T}$
- Dilepton invariant mass (m_{ll})
- Leading jet $p_{\rm T}$
- Subleading jet $p_{\rm T}$
- Dijet invariant mass (m_{jj})
- Lepton-jet centrality (ζ)

7.4.3 Results of the optimization

7.5 Cross section measurement

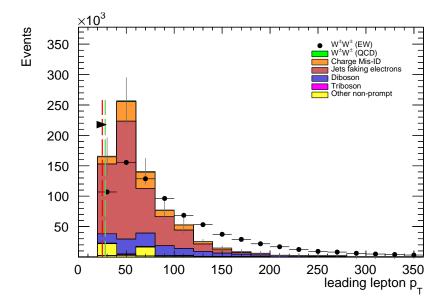


Figure 7.3: Leading lepton $p_{\rm T}$ distribution. The default and optimized cuts are represented by the red and green dashed lines, respectively. The $W^{\pm}W^{\pm}jj$ EWK signal (black points) is normalized to the same area as the sum of the backgrounds (colored histogram).

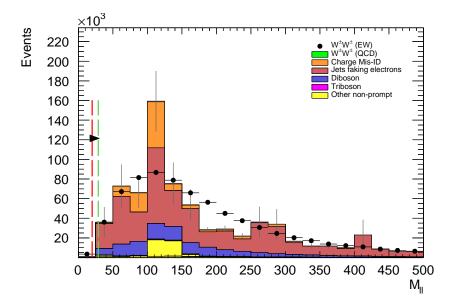


Figure 7.4: Dilepton invariant mass distribution. The default and optimized cuts are represented by the red and green dashed lines, respectively. The $W^{\pm}W^{\pm}jj$ EWK signal (black points) is normalized to the same area as the sum of the backgrounds (colored histogram).

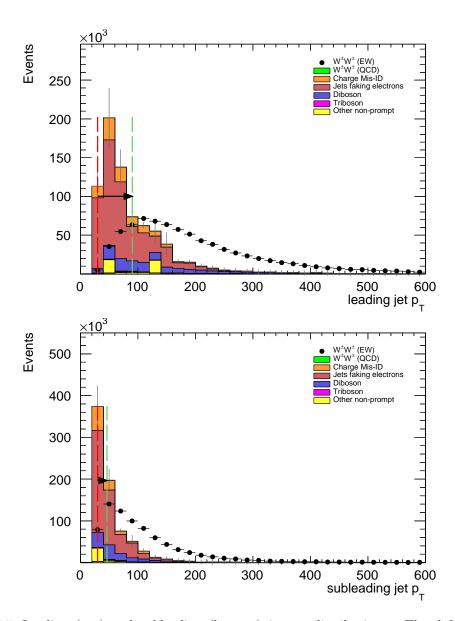


Figure 7.5: Leading (top) and subleading (bottom) jet $p_{\rm T}$ distributions. The default and optimized cuts are represented by the red and green dashed lines, respectively. The $W^\pm W^\pm jj$ EWK signal (black points) is normalized to the same area as the sum of the backgrounds (colored histogram).

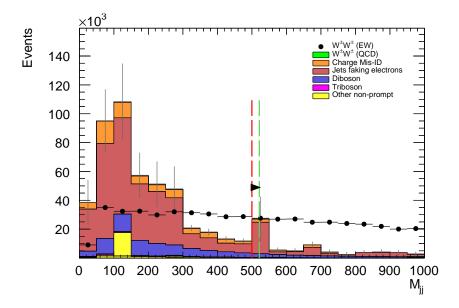


Figure 7.6: Dijet invariant mass distribution. The default and optimized cuts are represented by the red and green dashed lines, respectively. The $W^{\pm}W^{\pm}jj$ EWK signal (black points) is normalized to the same area as the sum of the backgrounds (colored histogram).

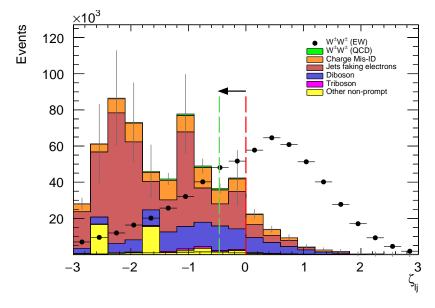


Figure 7.7: Lepton-jet centrality distribution. The default and optimized cuts are represented by the red and green dashed lines, respectively. The $W^{\pm}W^{\pm}jj$ EWK signal (black points) is normalized to the same area as the sum of the backgrounds (colored histogram).

Conclusion

265 Here's where you wrap it up.

266 Looking Ahead

263

267

Here's an example of how to have an "informal subsection".

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