

1st Year PhD Report

Using $Z \rightarrow \tau\tau$ events to calculate Tau ID
scale factors using high p_T Tau leptons

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Declaration

I hereby declare, that I am the sole author and composer of my thesis and that no other sources or learning aids, other than those listed, have been used. Furthermore, I declare that I have acknowledged the work of others by providing detailed references of said work.

I hereby also declare, that my Thesis has not been prepared for another examination or assignment, either wholly or excerpts thereof.

Place, Date

Signature

Abstract

foo bar

Contents

1	Introduction	1
2	Tau physics overview	3
2.1	The Tau Lepton	3
2.2	Lepton Universality	3
3	Analysis	5
3.1	The LHC and the ATLAS experiment	5
3.2	Tau Identification on the ATLAS detector	5
3.3	Monte Carlo Samples	5
3.4	The Collinear Approximation	5
3.5	Event Selection	5
4	Results	7
4.1	$\mu\tau$ Final state	7
4.2	$e\tau$ Final state	7
4.3	Monte Carlo and data discrepancies	7
5	Conclusions and prospects	9
	Bibliography	11

List of Figures

1	Tikz Example	6
2	Caption that appears in the figlist	10

List of Tables

1	Table caption	9
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List of Algorithms

1	Stochastic Gradient Descent: Neural Network	6
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1 Introduction

2 Tau physics overview

Give a brief overview of the work relevant for your thesis.

2.1 The Tau Lepton

2.2 Lepton Universality

3 Analysis

Explain the math and notation.

3.1 The LHC and the ATLAS experiment

3.2 Tau Identification on the ATLAS detector

3.3 Monte Carlo Samples

3.4 The Collinear Approximation

3.5 Event Selection



Figure 1: Use tikz to draw nice graphs!

Algorithm 1 Stochastic Gradient Descent: Neural Network

Create a mini batch of m samples $\mathbf{x}_0 \dots \mathbf{x}_{m-1}$

foreach sample \mathbf{x} **do**

$\mathbf{a}^{\mathbf{x},0} \leftarrow \mathbf{x}$

\triangleright Set input activation

foreach Layer $l \in \{1 \dots L-1\}$ **do**

\triangleright Forward pass

$\mathbf{z}^{\mathbf{x},l} \leftarrow \mathbf{W}^l \mathbf{a}^{\mathbf{x},l-1} + \mathbf{b}^l$

$\mathbf{a}^{\mathbf{x},l} \leftarrow \varphi(\mathbf{z}^{\mathbf{x},l})$

end for

$\delta^{\mathbf{x},L} \leftarrow \nabla_{\mathbf{a}} C_{\mathbf{x}} \odot \varphi'(\mathbf{z}^{\mathbf{x},L})$

\triangleright Compute error

foreach Layer $l \in L-1, L-2 \dots 2$ **do**

\triangleright Backpropagate error

$\delta^{\mathbf{x},l} \leftarrow ((\mathbf{W}^{l+1})^T \delta^{\mathbf{x},l+1}) \odot \varphi'(\mathbf{z}^{\mathbf{x},l})$

end for

end for

foreach $l \in L, L-1 \dots 2$ **do**

\triangleright

\triangleright Gradient descent

$\mathbf{W}^l \leftarrow \mathbf{W}^l - \frac{\eta}{m} \sum_{\mathbf{x}} \delta^{\mathbf{x},l} (\mathbf{a}^{\mathbf{x},l-1})^T$

$\mathbf{b}^l \leftarrow \mathbf{b}^l - \frac{\eta}{m} \sum_{\mathbf{x}} \delta^{\mathbf{x},l}$

end for

4 Results

The approach usually starts with the problem definition and continues with what you have done. Try to give an intuition first and describe everything with words and then be more formal like ‘Let g be ...’.

4.1 $\mu\tau$ Final state

4.2 $e\tau$ Final state

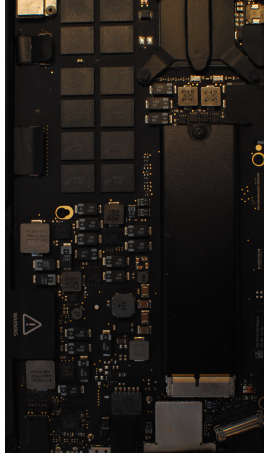
Start with a very short motivation why this is important. Then, as stated above, describe the problem with words before getting formal.

4.3 Monte Carlo and data discrepancies

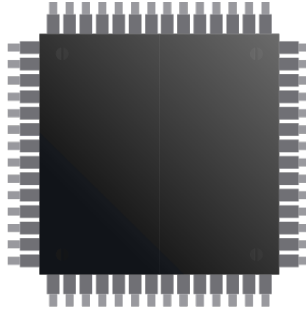
5 Conclusions and prospects

Type	Accuracy
A	82.47 ± 3.21
B	78.47 ± 2.43
C	84.30 ± 2.35
D	86.81 ± 3.01

Table 1: Table caption. foo bar...



(a) Some cool graphic



(b) Some cool related graphic

Figure 2: Caption that appears under the fig Lorem ipsum dolor sit amet, consectetur adipiscing elit. Ut purus elit, vestibulum ut, placerat ac, adipiscing vitae, felis. Curabitur dictum gravida mauris. Nam arcu libero, nonummy eget, consectetur id, vulputate a, magna. Donec vehicula augue eu neque. Pellentesque habitant morbi tristique senectus et netus et malesuada fames ac turpis egestas. Mauris ut leo. Cras viverra metus rhoncus sem. Nulla et lectus vestibulum urna fringilla ultrices. Phasellus eu tellus sit amet tortor gravida placerat. Integer sapien est, iaculis in, pretium quis, viverra ac, nunc. Praesent eget sem vel leo ultrices bibendum. Aenean faucibus. Morbi dolor nulla, malesuada eu, pulvinar at, mollis ac, nulla. Curabitur auctor semper nulla. Donec varius orci eget risus. Duis nibh mi, congue eu, accumsan eleifend, sagittis quis, diam. Duis eget orci sit amet orci dignissim rutrum.

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