Identification of Hadronically Decaying Tau Leptons at the ATLAS Detector Using Deep Neural Networks

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The Task

Distinction Between

- jets from hadronic tau decays (signal)
- jets from pure QCD processes (background)



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Hadronic Tau Decay



Figure: ATLAS Experiment © 2014 CERN

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Binary Classifier

- input: data points $\vec{x} \in \mathbb{R}^n$
- output: discriminating score $y \in [0; 1]$

Training

- training: iterative computation of the classifier's parameters
- training sample: data points with known results

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figures of merit: signal efficiency, background rejection

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Artificial Neurons



Single Layer Perceptrons



Figure: In binary classification, an SLP has only one active neuron.

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Single Layer Perceptrons



Multi Layer Perceptrons



Figure: Each hidden neuron takes up one part of the problem.

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Multi Layer Perceptrons



Figure: Solving the XOR problem with two hidden neurons.

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Deep Neural Networks



Figure: The first hidden layers transforms the input data.

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Problem

- How many layers do we need?
- How many neurons per layer do we need?
- What training algorithm works best?

Solution

- scan over all possible options
- train and evaluate an ANN for each configuration
- compare with current tau identification algorithm

What we used

- TMVA's MLP algorithm
- training algorithm: Backpropagation & BFGS algorithm
- benchmark: Boosted Decision Tree (BDT)

Training sample

- signal: Monte-Carlo generated
 - contains processes $W \to \tau \nu_{ au}$, $Z \to \tau au$, and $Z' \to au au$
- background: QCD jet data collected at ATLAS in the 2012 run

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Number of Neurons (Backprop, 1 Hidden Layer)



Network Depth (Backprop, 20 Neurons/Layer)



Training Methods (4 Hidden Layers @ 20 Neurons)



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Summary

- first attempt of NN-based tau identification
- comparison with current tau ID algorithm (BDT)
 - difference by a factor of ~ 2
- evaluation of performance under variation of
 - training method
 - neurons per layer
 - number of layers
- greatest influence: number of layers

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Outlook

- Is there a better choice of ID variables?
- Will even more layers improve the performance further?
- How do other training methods perform? (Quickprop, RProp, \dots)
- Where does overtraining begin?
- If we reach overtraining, can more sophisticated algorithms help?

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Backup

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Figure: Transfer Function $\varphi(x)$

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BDT Options

- NTrees=100
- DoBoostMonitor=True
- H=False
- UseYesNoLeaf=False
- nCuts=200
- PruneMethod=NoPruning
- BoostType=AdaBoost
- AdaBoostBeta=0.2
- MaxDepth=8
- V=False
- MinNodeSize=0.1
- SeparationType=GiniIndex

MLP Options

- RandomSeed=42
- NeuronType=sigmoid
- NCycles=200
- BPMode=sequential
- TrainingMethod=[...]
- HiddenLayers=[...]
- VarTransform=N
- ConvergenceTests=5
- ConvergenceImprove=1e-08

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Variables for 1-prong

- tau_calcVars_corrCentFrac
- tau_calcVars_corrFTrk
- tau_pi0_vistau_m
- tau_ptRatio
- tau_seedCalo_trkAvgDist
- tau_pi0_n
- tau_ipSigLeadTrk
- tau_seedCalo_wideTrk_n

Variables for 3-prong

- tau_calcVars_corrCentFrac
- tau_calcVars_corrFTrk
- tau_pi0_vistau_m
- tau_ptRatio
- tau_seedCalo_trkAvgDist
- tau_pi0_n
- tau_massTrkSys
- tau_seedCalo_dRmax
- tau_trFlightPathSig

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Signal Monte Carlo Samples

- $W \rightarrow \tau \nu_{\tau}$ (64k events)
- $Z \rightarrow au au$ (68k events)
- $Z' \rightarrow au au$, $m_{Z'} = 250 \, {
 m GeV}$ (62k events)
- Z'
 ightarrow au au , $m_{Z'} = 500 \, {
 m GeV}$ (62k events)
- $Z' \rightarrow \tau \tau$, $m_{Z'} = 1000 \text{ GeV}$ (61k events)

Background Samples

period D QCD jet data collected at ATLAS in the 2012 run (326k events)

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Network Depth (BFGS, 20 Neurons/Layer)



Training Methods (1 Hidden Layer, 20 Neurons)

