From pixel detectors to a search for new physics

IMRPS Recruitment Workshop

Tiziano Pauletto, 24th July 2024

arxiv.org/abs/astro-ph/0504097

A brief story My education and research experience

Education

- Obtained my bachelor's degree at the University of Trieste;
- Internship at INFN on the search for gamma counterparts to FRBs with Fermi LAT;
- BSc thesis on evaluation of PDF uncertainties in W boson's mass measurement at the LHC;
- Studying for my master's in Nuclear and Subnuclear Physics at the University of Trieste;
- Internship and MSc thesis on the characterization of a DNN for a search for new physics in $t\bar{t}$ final states in CMS.



Other research experiences

- DESY Summer Student 2023 in Hamburg: simulation of monolithic silicon pixel detectors;
- «Particle Physics to explore the Universe» 3 months INFN scholarship at CERN during this summer: working on track reconstruction in CMS's MTD and HGCAL.





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DESY Summer Student Simulations of monolithic CMOS pixel detectors

- Worked in the DESY Tangerine and ATLAS groups;
- Simulations of detector prototypes for use in future colliders;
- Used AllPix-Squared (based on GEANT4) for simulations and ROOT for data analysis;
- Goal: optimizing the detector's spatial resolution for future prototypes.









S. Spannagel

x (mm)

Master's thesis

Search for new physics in the top-antitop final state at CMS

- I'm working in a search for new physics of the CMS experiment;
- Analysis regarding Run II data collected at $\sqrt{s} = 13 TeV$;
- Top quark pair final state studied in semileptonic decays;
- Two considered topologies: resolved and boosted;
- Resolved: isolated lepton and Ak4 jets;
- Boosted: no requirement of isolated lepton, larger (Ak8) jets;
- Considered signals: Kaluza-Klein gluon, Z' as a dark sector-Standard Model mediator, Z' as a topcolor boson, additional Higgs bosons and ALPs.





D0 Collaboration, "Top Pair Branching Fractions"

Beyond Standard Model physics Particles that may decay to top quark pairs

Spin 1

- Kaluza Klein gluon g_{KK} postulated by the Randall-Sundrum model of extra dimensions, at various masses from 500 to 6000 GeV;
- Z' as a dark sector mediator, masses: 500 6000 *GeV*;
- Z' as leptophobic topcolor boson, with preferential coupling to top quarks. $\Gamma/m = 1,10,30\%$. Masses 400 9000 GeV;

Spin 0

- Scalar or pseudoscalar Higgs boson predicted by the 2 Higgs Doublet Model with widths $\Gamma/m = 2.5, 10, 25\%$. Masses ranging from 365 to 1000 GeV;
- Light Axion Like Particles (m < 100 GeV).



K. de Leo

The Deep Neural Network Architecture and goals of the work

- The analysis uses a Deep Neural Network for event classification;
- Architecture: two hidden layers, 512 neurons each;
- Input: 59 variables that describe each event;

Object	variable	
lepton	$p_T, \eta, \phi, \mathrm{E}$	
neutrino	p_T, ϕ	5 Ak4 jets 3 Ak8 jets
AK4 jets	N, p_T , η , ϕ , E, m, b-tag score	
AK8 jets	N, p_T , η , ϕ , E, m _{SD} , τ_{21} , τ_{32}	

- As output, score of event belonging to three classes: $t\bar{t}$, single top, and V+jets. Used to define signal and control regions;
- First goal: adding a class for signals and optimizing the performance of the DNN in identifying signal from background;
- Second goal: parametrizing the DNN in order to enhance sensitivity to new physics at masses not used in training.





How to assess the DNN's performance An example: confusion matrices



Confusion matrix for validation set, columns normalized



Confusion matrices are useful to study the DNN's performance in the classification task.

True label

Parametrization of the neural network Working principle

- The MC signals used in training are generated for particular signal masses;
- Computing resources constrain the range of masses that can be generated;
- Parametrized neural networks are introduced with the goal of enhancing the DNN's performance on signals not seen during training and making the model mass-agnostic.





arXiv:1601.07913

Parametrization of the neural network An example: the Z' sample

- The MC signals used in training are generated for particular signal masses;
- Computing resources constrain the range of masses that can be generated;
- Parametrized neural networks are introduced with the goal of enhancing the DNN's performance on signals not seen during training and making the model mass-agnostic.



$$F1 = 2 \frac{precision \cdot recall}{precision + recall}$$

F1 scores, comparing the DNNs trained on all masses with DNNs trained excluding the 2500,3000 GeV samples



F1 scores for the Z' dark matter samples

Thank you for your attention

NASA, ESA, and J. Lotz and the HFF Team (STScl)

Backup

NASA, ESA, and J. Lotz and the HFF Team (STScl)

Master's thesis Top quark production cross section



https://twiki.cern.ch/twiki/pub/LHCPhysics/TopPairCrossectionSqrtsHistory/tt_curve_topIhcwg_nov23.png

Master's thesis The kt algorithm

- If the smallest among the distances is d_{ij} with, i, j not being the beam axis, then combine the two entities in a single one;
- If the smallest distance is of the king d_{iB}, then call i a jet and remove it from the list of objects;
- Repeat the same procedure for all the objects.

$$d_{ij} = \min(k_{ti}^{2p}, k_{tj}^{2p}) \frac{\Delta_{ij}^2}{R^2},$$

 $d_{iB} = k_{ti}^{2p},$

- With p = 1 one obtains the k_t algorithm;
- p = 0 leands to the Cambridge-Aachen algorithm;
- Finally, p = 1 leads to the anti- k_t algorithm.



Master's thesis Output scores of the DNN trained on the Z' signal



(c) Output scores for the class V+jets.

(d) Output scores for the signal class.

1.0

1.0

Master's thesis Parametrized vs non-parametrized network

F1 scores, comparing the DNNs trained on all masses with DNNs trained excluding the 2500,3000 GeV samples



Master's thesis How to probe the new physics models



Master's thesis How to probe the new physics models

An observable sensitive to the spin of the mediator particle and can be used to probe new physics models is $\cos(\theta^*)$. Def. θ^* : angle between the $t\bar{t}$ system in the laboratory reference frame and the leptonically decaying top quark.





K. de Leo

Spatial resolution as function of the QDC resolution





Unexpected behaviour: spatial resolution gets worse with better QDC resolution

Residual distribution for different QDC resolutions



Residuals in the x direction

There are some precise ($\sim 1 \mu m$) track position reconstructions for 1-4 bit resolutions that do not show up at better QDC resolutions

Residual map for different QDC resolutions



Mean absolute deviation for in-pixel impact position: 1 bit

Mean absolute deviation for in-pixel impact position: 4 bits

0

Entries 99910

10

x%pitch [um]

20

18

16

14

12

10

 Hear absolute deviation for in-pixel impact position: 12 bits

Better track position reconstruction at the sides of the pixel for low QDC resolution



Residual map for different QDC resolutions - log scale

10

-10



Mean absolute deviation for in-pixel impact position: 1 bit

y%pitch [µm]

Mean absolute deviation for in-pixel impact position: 4 bits

Entries

99910

10

Better track position reconstruction at the sides of the pixel for low QDC resolution



23



'Monte Carlo Simulations of Detector Prototypes Designed in a 65 nm CMOS Imaging Process', Sara Ruiz Daza 24



- Plotting true track position vs reconstructed track position
- Fitting with a 5th order polynomial
- Afterwards all reconstructed track positions are recalculated using the fitted function





Gain of $\sim 0.2 \mu m$ in spatial resolution, but no dramatic change of it for different QDC resolutions. Now that is the expected behaviour of the detector: the n-gap design has a small fraction of events with cluster size > 1. Only those events gain in spatial resolution from the charge sharing across pixels. So the resolution's value is mostly due to the cluster size 1 events.