

Search for an additional Higgs boson with mass <110 GeV with the CMS experiment at the LHC 1

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<u>Search for an additional Higgs boson</u> with mass <110GeV with the CMS experiment at the LHC

Benjamin Massoteau









- This Higgs boson in the Standard Model
- Motivation for the search for an additional Higgs boson
- The CMS experiment
- ▶ The search for a new Higgs boson in the CMS experiment
- Discrimination between the Drell-Yan process and the diphoton decay
- Modelisation of the Drell-Yan process
- Conclusion

The Higgs boson

The Higgs mechanism is responsible for the spontaneous breaking of electroweak symmetry, allowing the W and Z bosons to acquire mass, as well as the fermions, thereby mixing their righthanded and left-handed components. The Higgs potential is as follows:

 $V(\phi) = \mu^2 \phi^{\dagger} \phi + \lambda (\phi^{\dagger} \phi)^2$

With $\mu^2 < 0$, the minimum of the potential is located at $v = \sqrt{-\frac{\mu^2}{\lambda}}$ This minimum is called the vacuum expectation value (vev). If we expand the Lagrangian around this vev, we notably find a mass term $\frac{m_h^2}{2}h^2$ with $m_h^2 = 2\mu^2$, reflecting the appearance of a scalar boson, a relic of this mechanism.



The Higgs boson discovery at the LHC

In 2012, the discovery of a scalar boson, compatible with the Higgs boson, was announced by the ATLAS and CMS experiments at CERN. This discovery was observed in the diphoton decay channel with a significance greater than 5σ , providing strong evidence of its existence. This marked a significant milestone in completing the Standard Model of particle physics, confirming the mechanism responsible for particle mass acquisition as predicted by the Higgs mechanism.



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Motivation for the search for an additional Higgs boson

The Standard Model, however, fails to predict and explain certain phenomena:

- •Dark matter, which is thought to constitute 24% of our universe.
- •Gravity at the quantum scale.
- •Neutrino oscillations.
- •The matter-antimatter asymmetry.
- ⇒ There must exist a theory beyond the Standard Model that can explain all observed and unexplained phenomena.

Certain models, such as two-Higgs-doublet models or supersymmetry, predict additional Higgs bosons with masses different from the one detected in 2012. An excess was notably detected around 95.4 GeV, measured with a local significance of 2.9 σ by CMS, which could correspond to an additional Higgs boson. Meanwhile, ATLAS measured a significance of 1.7 σ around the same mass <u>arXiv:2407.07546</u>

arXiv:2405.18149 (Accepted by PLB)





KMS experiment

B

KRAMPOIL OBUOZ .

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Krampouze Meson Spectrometer

<u>CMS experiment localisation</u>



The farthest point from CERN's main site (~20 minutes by car)

The CMS experiment



The CMS experiment





<u>The electromagnetic calorimeter</u>



The high energy resolution electromagnetic calorimeter (ECAL) measures the energy of photons and electrons. It is composed of PbWO₄ (lead tungstate) crystals.

Electron : Brernsstrahlung ; Photon : Pair production





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The particle cascade is then detected by photodetectors, allowing the reconstruction of the energy of the incident particle.

Low mass h->yy analysis

- Production mode : ggH, VBF, VH, et ttH
- Diphoton invariant mass :
 - Signal search range: 70-110 GeV
 - Background fiting range: 65-120 GeV
- Backgrounds :
 - Irreducible background producing γγ (QCD)
 - γ+jets, jets+jets
 - Drell-Yan : Z/γ*->ee





Problem of the Drell-Yan process reduction

To discriminate photons and electrons in the CMS experiment, both information from tracker and ECAL are used.

Electron : charged particle that leave an inner track

<u>Photon</u> : neutral particle that do not have such track

Some ineffiencies in the tracker misinterpret electrons as photons, especially coming from the Drell-Yan process.



Problem of the Drell-Yan process reduction

Pun 2 solution :

Integrated luminosity : L=132.2fb⁻¹

Linear cut in 2D between p_T (transversal momentum) and $\log(\sum p_T^2_{\gamma\gamma})$ (Sum of the square of the two transversal momenta of the tracks in the chosen vertex) such that : $\log(\Sigma p_{T\gamma\gamma}^2) \le 0.016 p_T + 6.0$

With the additional selection $N_{MatchedEle} = 0$

Thus, we keep 92% of the signal and keep 30% of the background.

Challenge for Run 3 :

- Data format have changed since Run 3 and do not have the variables $\log(\sum p_T^2_{\gamma\gamma})$ and $N_{MatchedEle}$
- We need the same results as in Run 2, or even better.

arXiv:2405.18149 (Accepted by PLB)



Here, electronIdx indicates whether a photon originates from an electron or not (in the latter case, the variable is set to -1). ⇒ Shown to be equivalent to N_{matched} = 0

PV_score corresponds to the sum of the transverse momenta of the clustered objects (Not only tracks).

We can try to do a linear cut.

With this method, the efficiency is : - Signal efficiency= 93%

- DY efficiency= 36,4%



<u>Neural Network</u>



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To discriminate the signal from our background, we're investigating using a neural network that can gives us the probability of an event to belong to one or another.

We have a training set and a test set of events that we label (signal = 1; background =0).

Each event will go into a unit called a neural, which is a matrix filled with scalars called weights.



<u>Neural Network</u>

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Every output of every neural has an activation function : Identity, ReLU, sigmoid, etc...

The ReLU function gives us some non linearity in the model :

ReLU(x)=max(0,x)

The sigmoid is used to have an output between 0 and 1, so our result can be interpreted as a probability.

$$\sigma(z) = \frac{1}{1 + e^{-z}}$$



The result is then compared to the training set using a loss function that has to be minimized.

<u>Neural Network</u>

The list of variables used comes from the photon kinematics (eta,phi,p_T) from the most two energetics photons (lead and sublead), informations from the primary vertex (PV_*) and from energy deposit in the ECAL and the tracker.



Neural Network

Loss function



<u>Selection</u>

After training the model, we can calculate the importance of the selected variables by ranking them based on the value of their gradient when a small perturbation is introduced.

We can also calculate the efficiency by performing a selection associated with the score of each event, i.e., the probability that it originates from the signal or background noise. The best possible cut is 0.90.

Eff_s = 94% Eff_bkg = 6%

But this is only true for a signal at 90GeV

Work still on going...





Background modeling

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The number of Monte Carlo events is not enough to do a proper modeling so we are investigating using a data augmentation algorithm. The technique selected is called a variational autoencoder.

The idea is to encode a distribution into a latent space and then decode it to generate a similar one with greater statistics.



Diphoton BDT

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A diphoton BDT (boosted decision tree) have been trained to discriminate prompt diphoton events from the QCD events (γ +jet,jet+jet).

The score is then used to divide photon events into 4 classes such that we have the best efficiency for all of them.



Background modeling challenge



Data Augmentation example : Mass

24 The number of events are the same for both trained and generated events

+ Original Data

Generated Data

Epoch = 100



Epoch = 300



Epoch = 500200



Epoch = 1000









Data Augmentation example : BDT

Epoch = 1500

25







Epoch = 3000







Epoch = 2500



Data Augmentation

$$F_X(x) = P(X \le x)$$

KS Stat = $\max_{bin}(F_{reel}, F_{VAE})$



CDF of the mass distribution

KS Statistics = 0.03 P-value = 0.23

CDF of the BDT distribution



KS Statistics = 0.13P-value = 10^{-22}

CDF for each classes











80

90

100

Value

CDF of Original

70

KS Stat = 0.050P-Value = 0.25







- ► The strategy of the analysis is still on going :
 - A neural network is being currently investigated to discriminate h->γγ events from Z->ee events
 - The VAE has to be optimized and then fit to a function to see if the results are better than without
 - The VAE has to be test with mass, diphoton BDT and DY suppression neural network score

THANKS FOR YOU ATTENTION

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Back ups

Plot ATLAS for Run 2 search of low mass Higgs Boson

With the full Run 2 data, ATLAS has found an excess around 95.4GeV with a significance of 1.7 σ . Which is less than CMS (2.8 σ). Giving us a combined excess of 3.8 σ .

arXiv:2407.07546



Neural Network

For each neural, the corresponding result is $z = \sum_i f(w_i x_i + b)$, we do this for every neural for each layers. This is called the forward propagation.

Then we compute the loss function. For a classifier, it corresponds to the binary cross entropy : $BCE(x) = -\sum y_n \ln(x_n) + (1 - y_n) \ln(1 - x_n)$. Then, we perform a backpropagation, corresponding to the correction of the weights such that the loss function can be minimized.

