Higgs pair production at the FCC-hh

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Introduction

- Previous studies using a BDT were developed in 2022 (<u>see presentation at Higgs pair by Matt</u> <u>Sullivan</u>)
 - Results taking into account both $\tau_{\rm L}\tau_{\rm H}$ and $\tau_{\rm H}\,\tau_{\rm H}$
 - Very good sensitivity, comparable with publishec _ studies (<u>https://arxiv.org/pdf/2004.03505</u>)



This work: implement graph neural networks

- GNN pipeline from Alessio Devoto (PhD Computer Scientist, University of Rome Sapienza)
- · Graph for each event, each object is a node
- · Fully connected, each node has several features
- Different models tested (GCN, GAT)
- Systematic evaluation of performance based on relevant metrics (S vs B separation, AUC)
- Inputs and samples using official samples (EDM4HEP format) and ntuples generated with FCC analysis starterkit (same as linked in Matt's slides above)

FCC simulation

- Baseline FCC-hh detector response simulated using Delphes (v4) parameterisation
- Lepton (e, μ) and photon reconstruction employs parameterised reco/ID efficency & resolution effects
- Jet reconstruction uses Anti-kT algorithm with R = 0.4
- Object isolation calculated using cone of R = 0.3
- b-tagging, c-tagging and τ -tagging efficiency parameterised in pT, η





FCCAnalysis framework

- Common RDataFrame analysis framework developed for FCC physics studies: FCCAnalyses

- Common C++ analysers, analysis-specific Python config & analysis:

- See C. Helsens talk for example workflow
- FCC analysis starterkit
- Inputs to analyses are produced in EDM4HEP format:
 - All available MC listed here

- Efficient analysis possible with handful of scripts



HH at FCC

- Numerous existing studies on HH at FCC-hh:

- HH production (b⁻bb⁻b, b⁻bττ, b⁻bγγ)
- HH + jet production (boosted b⁻bb⁻b, b⁻btt, resolved b⁻btt)
- Combination of resolved channels has expected $\delta\mu$ of 2.4-5.1%, $\delta\kappa\lambda$ of 3.4-7.8%
- Boosted b⁻btt can constrain κλ to within 8% alone!
- What can be improved upon?



arXiv 2004.03505

bbtautau channel

- Focus on HH \rightarrow bbtt channel
- Use more modern MVA tools to improve S/B:
 - GNN and GraphTransformers
- Use latest FCC-hh simulated samples with more complete background estimation:
 - Top backgrounds: t⁻t, single top (s-/t-channel),

t⁻tV, t⁻ tVV

- Single Higgs backgrounds: ggF, VBF ,t⁻tH, VH
- Continuum backgrounds: QCD+EW (e.g. pp

 $\rightarrow b^-bZ /\gamma *)$, EW (e.g.pp $\rightarrow HZ /\gamma *)$

	bb	ww	ττ	ZZ	ΥY
bb	34%				
ww	25%	4.6%			
ττ	7.3%	2.7%	0.39%		
zz	3.1%	1.1%	0.33%	0.069%	
YY	0.26%	0.10%	0.028%	0.012%	0.0005%

Process	σ [pb]			
ggF HH	1.224 (NNLO _{FT})			
Inclusive $t\overline{t}$	35000 (NNLO)			

preliminary selections

- Apply loose topological and kinematic cuts:
 - b⁻bt ℓ th : 2 b-jets, exactly 1 e/ μ and exactly 1 hadronic t (OS)
 - b⁻bтh тh : 2 b-jets, exactly 2 hadronic т (OS), lepton veto
- Overlap removal prioritises taus over b-jets

$ au_\ell au_{ m h}$		$ au_{ m h} au_{ m h}$			
Selection	Efficiency	Selection	Efficiency		
N(b-jets) = 2	0.45	-	-		
$b ext{-jet} p_{\mathrm{T}} > 30$	0.40	-	-		
$\mathit{N}(e,\mu)=1$	0.07	$N(e,\mu)=0$	0.19		
$N(au_h)=1$	0.03	$N(au_h) = 2$	0.06		
$ au_\ell au_h \ OS$	0.03	$ au_h au_h$ OS	0.05		

GNN selection

GNN pipeline from Alessio Devoto (PhD computer Scientist, University of Rome Sapienza)

- Graph for each event, each object is a node
- · Fully connected, each node has several features
- Different models tested (GCN, GAT)



GNN performance





Additional feature nodes

Add complex reconstructed kinematic variables

- b-jet pairs invariant mass
- tau-lepton invariant mass
- radial distances among b and tau objects and ETMiss centrality as in ATLAS di-Higgs studies

$$E_T^{miss}centrality = \frac{(x+y)^2}{\sqrt{x^2+y^2}}$$

$$x = rac{\sin(\phi_{MET} - \phi_{ au})}{\sin(\phi_{\ell} - \phi_{ au})}$$

$$y = rac{\sin(\phi_\ell - \phi_{MET})}{\sin(\phi_\ell - \phi_ au)}$$



GNN improved performance





Additional di-higgs constraints



- Adding m_hh and dphi_hh helps to further improve performance
- Possibility to not use these variables in the inputs to the GNN but use only for differential cross section measurements

the network is good enough without having to use the di-higgs system as constrain





Calculating significance

Calculate signal significance in NNoutput bins:

 $Z = N_s / \sqrt{N_b + (N_b \sigma_b)^2}$

with a signal and background scaled to 30/ab

Next step is to compare with previous BDT study and HH+jet study (Add per-bin significance in quadrature to get final estimate)

> Old BDT study: Significance Z = 5.7σ for $\kappa\lambda$ = 1: (2.9σ b bt{th, 4.9σ b bthth)



Summary

- First estimate of sensitivity show a significance similar to BDT-based results
 - Vanilla GNN tested so far... full optimisation is ongoing
- Limited by MC statistics, so next steps is to evaluate sensitivity with full stat ttbar and add fully hadronic channel
- Explore the had-had channel
 - How should we treat fakes?
- Once the GNN are finalised, define the full analysis strategy
 - Differential cross-section?
 - k_lambda fit?

BACKUP

Test	epochs	Batch size	Scheduler ?	LR	Weight decay	Hidden layers	Hidden channels	model	Traina bles?	Norma lised?	Complex features
1	100	500	No	0.001	5e-4	2	18	GCN	No	No	None
2	100	500	No	0.01	5e-4	3	50	GCN	No	No	None
3	100	500	No	0.01	5e-4	3	50	GAT	No	No	None
4	100	500	Yes	1e-7	1e-5	3	50	GCN	No	Yes	None
5	100	500	Yes	1e-7	1e-5	3	50	GAT	No	Yes	None
6	100	500	Yes	1e-7	1e-5	3	50	GAT	Yes	Yes	None
7	100	500	Yes	1e-7	1e-5	3	50	GAT	No	Yes	mbb
8	100	500	Yes	1e-7	1e-5	3	50	GAT	No	Yes	"+dRbb
9	100	500	Yes	1e-7	1e-5	3	50	GAT	No	Yes	"+mtt
10	100	500	Yes	1e-7	1e-5	3	50	GAT	No	Yes	"+dRtt
11	100	500	Yes	1e-7	1e-5	3	50	GAT	No	Yes	"+ Cmet
12	100	500	Yes	1e-7	1e-5	3	50	GAT	No	Yes	"+ dpT
13	100	500	Yes	1e-7	1e-5	3	50	GAT	No	Yes	" + transverse mass
14	100	500	Yes	1e-7	1e-5	3	50	GAT	No	Yes	" + mhh
15	100	500	Yes	1e-7	1e-5	3	50	GAT	No	Yes	"+ dPhi_hh



LOSS AND ACCURACY - EDITING GNN PARAMETERS

LOSS AND ACCURACY - ADDING COMPLEX FEATURES



OUTPUT DISTRIBUTIONS – GNN PARAMETERS

Test 1



Test 2: hidden channels =

Test 3: Extra layer

0.5

0.6

background

0.8

signal

OUTPUT DISTRIBUTIONS



OUTPUT DISTRIBUTIONS

10¹

10⁰

 10^{-1}

Frequency Density

Test 14: + mhh Test 12: +dpT Test 13: + transverse mass Model Output Distribution on Validation Dataset Model Output Distribution on Validation Dataset Model Output Distribution on Validation Dataset background background background signal signal signal Frequency Density 10¹ Frequency Density 10⁰ 10^{-1} 10-1 0.0 0.2 0.4 0.6 Predicted Probabilities 0.8 1.0 0.0 0.2 0.4 0.6 0.8 1.0 0.0 0.2 0.4 Predicted Probabilities Predicted Probabilities Model Output Distribution on Validation Dataset background signal



0.6

0.8

1.0





ROC CURVE



ROC CURVE



FEATURE LEARNING

- Cut data above and below GNN score of 0.7.
- Plotted complex variables for each iteration.
- What GNN gives high and low probabilities gives indication of how and what it is learning.

INVARIANT MASS OF B-JETS



INVARIANT MASS OF B-JETS



DISTANCE BETWEEN B-JETS



INVARIANT MASS OF LEPTONS



DISTANCE BETWEEN LEPTONS



CENTRALITY OF MET

DIFFERENCE IN PT VALUES (LEPTONS)

DIFFERENCE IN PT VALUES (LEPTONS)

Test 12: +dpT

Test 13: +transverse mass

dpT_T

TRANSVERSE MASS

TRANSVERSE MASS

Test 12: +dpT Test 13: + transverse mass Test 14: + mhh 40 Signal (score < 0.7) Signal (score < 0.7) Signal (score < 0.7) 35 Signal (score > 0.7) Signal (score > 0.7) 35 35 Signal (score > 0.7) Background (score < 0.7) Background (score < 0.7) Background (score < 0.7) 30 Background (score > 0.7) 30 30 Background (score > 0.7) Background (score > 0.7) 25 20 15 Density 20-15 25 Density 12 12 background signal 0 10 10 10 5 0.00 0.05 0.10 0.15 0.20 0.25 0.30 0.35 0.40 0.00 0.05 0.00 0.10 0.15 0.20 0.25 0.30 0.35 0.40 0.05 0.10 0.15 0.20 0.25 0.30 0.35 0.40 5 transverse mass transverse mass transverse mass 6 10 Signal (score < 0.7) 35 Signal (score > 0.7) Background (score < 0.7) 30 0.00 0.05 0.10 0.15 0.20 0.25 0.30 0.35 0.40 Background (score > 0.7) transverse mass 25 20 15 10 0.00 0.05 0.10 0.15 0.20 0.25 0.30 0.35 0.40 transverse mass

Test 15: + dPhi_hh

INVARIANT MASS OF TWO HIGGS

INVARIANT MASS OF TWO HIGGS

Test 12: +dpT

Test 13: + transverse mass

Test 14: + mhh

Test 15: + dPhi_hh

DIFFERENCE IN PHI OF TWO HIGGS

DIFFERENCE IN PHI OF TWO HIGGS

Test 12: +dpT

Test 13: + transverse mass

Test 14: + mhh

Test 15: + dPhi_hh