Cornering charming Higgs decays

Joseph Walker, Frank Krauss

Machine Learning for Jets Physics 2020 New York University 17th January 2020



Contents

- Why look at H -> cc
- Processes considered
- Data collection
- Observables, Images and Particle Flows
- Training case study
- FIFO Rivet <-> Python
- Network Voting
- Results and outlook

Can we pin down the yukawa coupling for light fermions?

$$\kappa_c = rac{y}{y_{SM}}$$

Current upper bound from CMS [1]

$$rac{\sigma(VH)Br(H o car{c})}{\sigma_{SM}(VH)Br_{SM}(H o car{c})} = \kappa_c^2 < 70(37^{+16}_{-10})$$

Charming higgs decays dominated by huge backgrounds

[1] M. Tanabashi et al. (Particle Data Group), Phys. Rev. D 98, 030001 (2018) and 2019 update 6th December, 2019

17/01/2020

Signals

We consider three signal processes



This gives us 3 channels;

0 Isolated Leptons 1 Isolated Leptons

2 Isolated Leptons

Signals

We consider three signal processes



This gives us 3 channels;

0 Isolated Leptons 1 Isolated Leptons

2 Isolated Leptons

Using fat jets as a gateway to background elimination

Backgrounds



In total 19 background runcards are used;

- Higgs couplings turned off
- MEPS(2)
- Decay of one vector boson is enforced in the hard matrix element

Data

Data is generated using Sherpa 2.7.7 at MEPS(2) for a selection of Backgrounds. Analysis performed by Rivet 2.7.0 and Fastjets 3.3.2

Many Runcards are used to boost the efficiency of backgrounds being generated.

We find fatjets with R=1 with the Anti-Kt algorithm with pT > 250 GeV

We apply a set of simple cuts to identify candidate fatjets

All surviving events are analysed and three data structures are saved;

- 1. Observables [2]
- 2. Jet Images [3]
- 3. Particle Flows [4]

[2] arXiv:1712.03634 & arXiv:1106.3076 [3] arXiv:1612.01551 & arXiv:1407.5675
 [4] arXiv:1810.05165v2



Z

Cutflows

We have 3 cut flows / 3 channels.

- 0. Eta and pT cuts on event
- A Missing Transverse Energy condition 0: < 10 GeV 1: > 80 GeV 2: < 10 GeV
- 2. Demand N isolated leptons
- 3. Demand at least 1 R = 1.0 fatjet, pT > 250 GeV
- 4. Pick fatjet with suitable properties
 - a. C tagged
 - b. |JetMass 125 GeV| < 30 GeV
- Demand 2 R = 0.4 jets OR Demand Isolated leptons and cadidate jet back-to-back.

Cuts based on arXiv:1912.01662





17/01/2020

Chosen Processes

We collect data from any process with an expectation of surviving events for $\int \mathcal{L} dt = 3000 f b^{-1}$ **0** Isolated Leptons 1 Isolated Leptons 2 Isolated Leptons

- 1. VBF (11%)
- 2. WZ[qq] (0.90%)
- 3. WW[qq] (0.14%)
- 4. t[bw]t[bw] (0.017%)
- 5. WZ[II] (0.012%)
- W[qq] (0.0034%) 6.
- 7. Z[II] (0.0002%)
- Z[qq] (0.0001%) 8.

These runcards are boosted to generate balanced classes

- W Higgs-Strahlung (10%) 1.
- t[bw]t[bw] (0.016%) 2.
- WW[lv] (0.0098%) 3.
- 4. W[lv]Z (0.0057%)
- W[lv] (0.0006%) 5.

- Z Higgs-Strahlung (17%) 1.
- 2. WZ[II] (0.16%)
- 3. Z[II] (0.0020%)
- Z[qq] (0.0020%) 4.



Chosen Processes - Cross sections

We collect data from any process with an expectation of surviving events for $\int \mathcal{L} dt = 3000 fb^{-1}$ 0 Isolated Leptons 1 Isolated Leptons 2 Isolated Leptons

- 1. VBF 0.19(4) pb
- 2. WZ[qq] 20(1) pb
- 3. WW[qq] 0.12(2) nb
- 4. t[bw]t[bw] 5.4(7) pb
- 5. WZ[II] 1.8(1) pb
- 6. W[qq] 90(5) nb
- 7. Z[II] 3.3(1) nb
- 8. Z[qq] 27(2) nb

 $\frac{S}{B}$

- W Higgs-Strahlung 1. 0.008(2) pb
- 2. t[bw]t[bw] 5.4(7) pb
- 3. WW[lv] 31(4) pb
- 4. W[lv]Z 2.9(3) pb
- 5. W[lv] 29(2) nb

- 1. Z Higgs-Strahlung 0.00123(4) pb
- 2. WZ[II] 1.8(1) pb
- 3. Z[II] 3.3(1) nb
- 4. Z[qq] 27(2) nb

 $rac{S}{R}\simeq 10^{-7}$

$$\simeq 10^{-6}$$
 $rac{S}{B}\simeq 10^{-7}$

LO Cross sections!









-0.2

Create RGB images

red ~ Energy green ~ Perpendicular momentum to jet momentum blue ~ Charge particle multiplicity

Jet axis lined up with (0,0) pixel Sum up variables in each bin then rescale

~15k Trainable Parameters

A mean 'background image'





Particles in jet ordered in decreasing energy

Form a sequence of 10 particles to be fed into a particle flow network.

1. $\Delta \eta$

 \mathbf{a}

A /

| Ζ. | $\Delta \phi$ | | DEta | DDbj | log(nt) | log(E) | log(nt/jotnt) | log(E/iotE) | DD |
|----|---------------------|----|-------------------|-----------|--------------------|--------------------|------------------------|------------------------|----------|
| 3 | $log(p_T)$ | 0 | 0.012631 | -0.054196 | 5.04692 | 5.57067 | -0.590013 | -0.570963 | 0.055648 |
| J. | $J(\mathbf{r} 1)$ | 1 | 0.334012 | -0.252075 | 2.35535 | 3.09847 | -3.281580 | -3.043170 | 0.418456 |
| 4. | log(E) | 2 | 0.231418 | 0.024186 | 2.67646 | 2.97307 | -2.960470 | -3.168560 | 0.232679 |
| _ | | 3 | 0.208372 | 0.038378 | 2.50181 | 2.81440 | -3.135130 | -3.327240 | 0.211877 |
| 5. | $log(p_T/p_{Tjet})$ | 4 | 0.349394 | -0.149882 | 2.02044 | 2.77714 | -3.616490 | -3.364500 | 0.380186 |
| 6. | $log(E/E_{jet})$ | 56 | 0.224771 0.117750 | 0.326364 | 2.24209 1.88444 | 2.54319 2.44333 | -3.394850 -3.752490 | -3.598450 -3.698310 | 0.396277 |
| | • | | | | | | | | |

- 7. ΔR
- 8. PID

~1.6k Trainable Parameters

The pipeline

- 1. Determine appropriate backgrounds for each channel
- 2. Boost events to give balanced classes ~10k samples
- 3. Normalise and rescale data
- 4. Learn in tensorflow



Observables



Observables







Sherpa -> Rivet Analysis

<-->

Python

| _ reduce wax for 5 7 1 1 m. [6 6] for a | υ./15086 (eps = 0.001) | Ims [0.97180384, 1] |
|--|--|----------------------|
| Read in channels from directory : Results | Flow [0.9024279, 1] | |
| | Obs [0.81896096, 1] | |
| 2_3_ <u>J_J_</u> W-[V_V]_J : 4244.04 pD +- (| Ims [0.96480846, 1] | |
| A coduce may for 2 2 i i U [0 0] i i | = 0.502677 (-0.5 - 0.001) | Flow [0.91917694, 1] |
| | 10 0.595677 (eps = 0.001) | Obs [0.99326575, 1] |
| - SHEPPA generates events with the follo | Ims [0.9584786, 1] | |
| Sheki'A generates events with the rotte | | FLow [0.89479154, 1] |
| Perturbative · Signal Processes | | Obs [0.994/601, 1] |
| Perturbative : Hard Decays | | Ims [0.9261591, 1] |
| Perturbative : let Evolution:CSS | | FLOW [0.8904244, 1] |
| Perturbative : Lepton ES OED Correc | tions:Photons | Obs [0.9974905, 1] |
| Perturbative : Multiple Interaction | | Ims [0.9860474, 1] |
| Perturbative · Minimum Bias:Off | 13:10116 | Flow [0.8871511, 1] |
| Hadsopization : Roam Dompants | | Obs [0.7706096, 1] |
| Hadronization : Hadronization: Abadi | - | Ims [0.9582879, 1] |
| | | Flow [0.9047799, 1] |
| | | Obs [0.9870056, 1] |
| Applysis Divet | | Ims [0.88494724, 1] |
| Analysis : River | | Flow [0.49075803, 0] |
| Divot Applycic Upodloct MADN Applycic / | TatlateQuest is upualidated, be caseful it ma | Obs [0.99569356, 1] |
| Rivet.Analysis.Handler: WARN Analysis i | -atjetswiep is unvalidated: de careful, it ma | Ims [0.939048, 1] |
| y de droken: Divet Anglusia Unadlana UADN, Anglusia U | | Flow [0.8998179, 1] |
| Rivet.Analysis.Handler: WARN Analysis I | -atjetsilep' is unvalidated: de carerul, it ma | Obs [0.99020326, 1] |
| y de droken! | | Ims [0.94132274, 1] |
| Rivet.Analysis.Handler: WARN Analysis I | -atjetszlep' is unvalidated: be careful, it ma | Flow [0.91626483, 1] |
| y be broken! | | Obs [0.9833592, 1] |
| # | | Ims [0.949657, 1] |
| # FastJet_release | 2 3.3.1 | Flow [0.9092387, 1] |
| # M. Cacciari, G.P. Salar | n and G. Soyez | Obs [0.99601537, 1] |
| # A software package for jet finding | Ims [0.9835242, 1] | |
| # http://fastje | et.fr | Flow [0.9188933, 1] |
| # | | Obs [0.42806774. 0] |
| <pre># Please cite EPJC72(2012)1896 [arXiv:11]</pre> | Ims [0.20386374, 0] | |
| # for scientific work and optionally PLB0 | Flow [0.8513089, 1] | |
| # | | Obs [0.9964684.1] |
| <pre># FastJet is provided without warranty up</pre> | Ims [0.97812784, 1] | |
| # It uses T. Chan's closest pair algorith | Flow [0.9187265, 1] | |
| # and 3rd party plugin jet algorithms. Se | ee COPYING file for details. | Obs [0.9929892, 1] |
| # | | Ims [0.9795156, 1] |
| Event 200 (1s elapsed / 5s left) -> I | Flow [0.90042114, 1] | |
| XS = 107484 pb +- (10097 pb = 9.39 %) | | |
| | | |

FIFO



Voting

Each network returns a probability for each class.



We have a naive voting system, but can we do better?

Voting

The networks are providing complementary information! More classifying power!



Voting

The networks are providing complementary information! More classifying power!



Only a few 1K data points (for now!)

Time to learn

Learn how to vote!



Combined networks give superior killing power than individual networks!



 $rac{\sigma(VH)Br(H
ightarrow car{c})}{\sigma_{SM}(VH)Br_{SM}(H
ightarrow car{c})}=\kappa_c^2$ $Br_{SM}(H o car{c}) = 2.89 imes 10^{-2} \, {}^{+5.5\%}_{-2.0\%}$ 1σ 95% confidence Z Higgs-Strahlung ML W Higgs-Strahlung ML X VBF ML × Z Higgs-Strahlung W Higgs-Strahlung VBF -2020 0 40 κ_c at $\int \mathcal{L} dt = 300 f b^{-1}$





Key points

- 1. A proof of concept
- 2. Promising results
- 3. Pipeline ready to go!

Considerations

- 1. This was only a LO study
- 2. Possibility to consider more backgrounds
- 3. Detector simulation
- 4. More data!



Backups

- 1. Network Architectures
- 2. Chosen Observables
- 3. Distributions

Architectures



Chosen Observables



A good observable: Jet Mass

A bad observable: Jet Rapidity

Chosen Observables



A bad observable: Jet Rapidity

A good observable: Jet Mass

17/01/2020

Distributions

Sanity checks - 2 Isolated lepton channel Candidate fatjet distributions

Leptons invariant mass

