

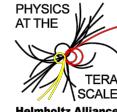


Fitting: Session 3

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Statistical introduction

Normal Distribution

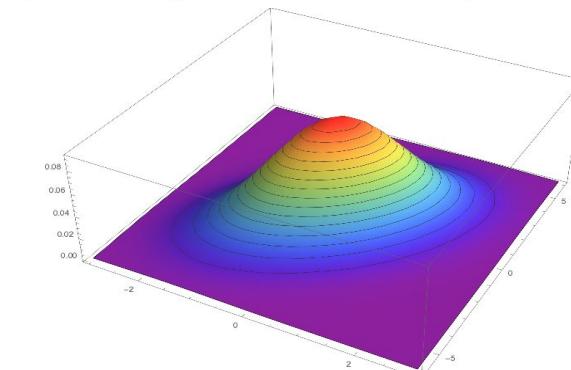
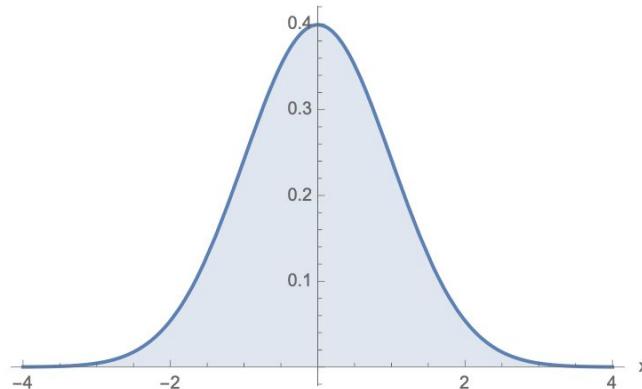


Continuos distributions

$$\mathcal{P}(x|\mu, \sigma) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$\mathbb{E}[x] = \mu$$

$$\mathbb{V}[x] = \sigma^2$$



ChiSquare Distribution



- Chi Square distribution
 - sum of square normal distribution

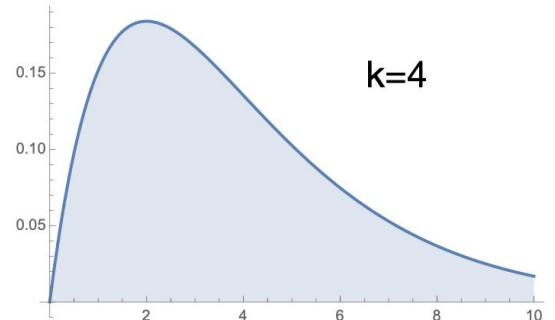
$$X_i \sim \mathcal{N}(0, 1)$$

$$Q = \sum_{i=0}^k X_i^2$$

k is the number
of degrees of freedom

$$\mathbb{E}[x] = k$$

$$\mathbb{V}[x] = 2k$$



Wilks' theorem



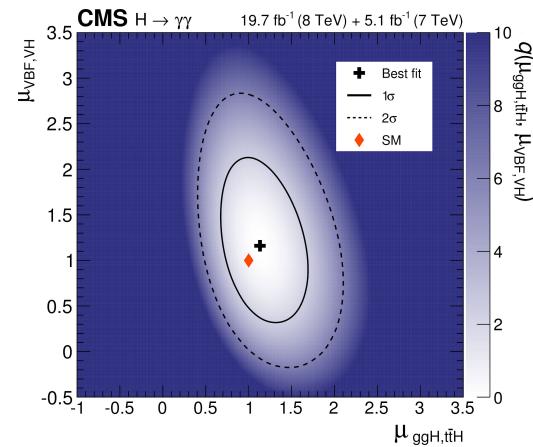
Likelihood ratio are often used in statistical tests:

- a) H_0 is true
- b) H_0 and H_1 are nested
- c) Params for H_1 , H_0 are well defined, and not on boundary
- d) Data is asymptotic, i.e. the sample size approaches to infinity

Then the $-2 \log \Lambda$ is distributed as a χ^2 distribution with N degrees of freedom (where N is the difference in number of parameters between H_1 and H_0)

and
$$\Lambda = \frac{L(\vec{\alpha})}{L(\hat{\vec{\alpha}})}$$

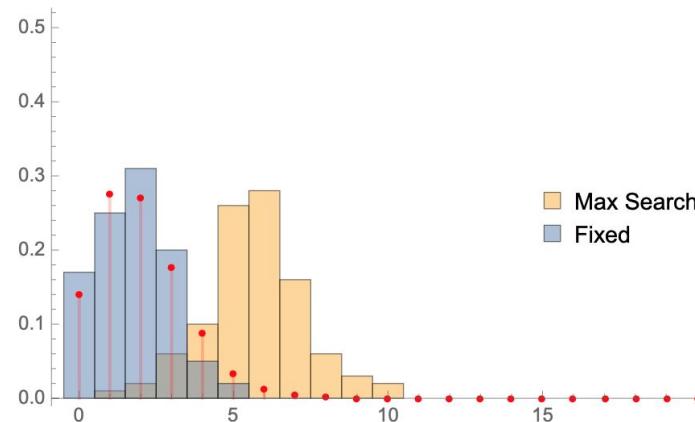
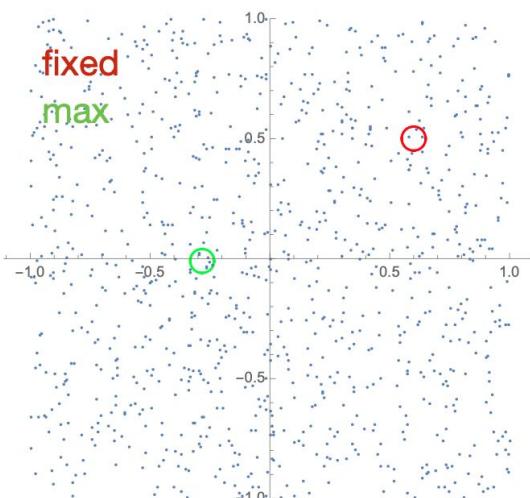
	68.3%	95%
1 ndf	1	3.84
2 ndf	2.23	5.99



A word on LEE



- Look Elsewhere Effect (LEE) appears when you “cherry-pick” statistical results
 - Maximum finding, ecc..
- Very difficult to quantify
 - especially with nuisances, ecc..
 - use toys or LogL crossings
- 1000 points in $[-1,1]^2$
- $r_0 = 0.05 \rightarrow$ Poisson with mean ~ 2
- how common is to observe 7?





Hands On

Objective of Hands On



What you will see today:

Material: <https://github.com/amarini/Prefit2020>

- MultiSignal Models
- Extract EFT couplings from STXS-like measurements



Part 1

Parametrization production



Look for available parametrization. We will use these ones
 (<https://cds.cern.ch/record/2706103/files/HIG-19-005-pas.pdf>). Stage 0.

STXS region (stage 0)	A_j
$gg \rightarrow H$	$8.73 \times 10^3 c_G$
$qq \rightarrow Hqq$	$9.02 c_{WW} + 0.6 c_B - 0.797 c_{HW} + 0.399 c_A$
$qq \rightarrow H\ell\nu$	$42.5 c_{WW} + 19.9 c_{HW}$
$qq \rightarrow H\ell\ell$	$36.6 c_{WW} + 10.5 c_B + 15 c_{HW} + 5.14 c_A$
$gg/\bar{q}q \rightarrow ttH$	$2.95 c_u + 115 c_G$

Table 10: A_j coefficients for the STXS stage 0 bins.

STXS region (stage 0)	B_{jk}
$gg \rightarrow H$	$1.95 \times 10^7 c_G^2$
$qq \rightarrow Hqq$	$171 c_{WW}^2 + 3.42 c_B^2 + 114 c_{HW}^2 + 0.874 c_A^2 + 23.1 c_{WW} c_B + 233 c_{WW} c_{HW} + 6.22 c_{WW} c_A + 15.3 c_B c_{HW} + 2.02 c_B c_A + 0.681 c_{HW} c_A$
$qq \rightarrow H\ell\nu$	$912 c_{WW}^2 + 558 c_{HW}^2 + 1.3 \times 10^3 c_{WW} c_{HW}$
$qq \rightarrow H\ell\ell$	$602 c_{WW}^2 + 51.7 c_B^2 + 321 c_{HW}^2 + 10.7 c_A^2 + 350 c_{WW} c_B + 772 c_{WW} c_{HW} + 102 c_{WW} c_A + 227 c_B c_{HW} + 31.4 c_B c_A + 29.7 c_{HW} c_A$
$gg/\bar{q}q \rightarrow ttH$	$2.14 c_u^2 + 6.13 c_{WW}^2 + 1 c_B^2 + 5.87 c_{HW}^2 + 2.97 \times 10^4 c_G^2 + 167 c_u c_G - 0.31 c_{WW} c_B + 11.9 c_{WW} c_{HW} - 0.318 c_B c_{HW}$

Table 11: B_{jk} coefficients for the STXS stage 0 bins.

Parametrization decay



And these ones (for the decay) <https://cds.cern.ch/record/2673969>

Partial width	$\sum_i A_i c_i$
$H \rightarrow b\bar{b}$	$-1.0cH + 3.0cd$
$H \rightarrow WW^* \rightarrow l\nu l\nu$	$10cWW + 3.7cHW + 2.2cpHL$
$H \rightarrow ZZ^* \rightarrow 4l$	$55cWW + 13cB + 15cHW + 4.6cHB + 0.018c_\gamma + 2.0cHL + 2.0cpHL + 0.027cHe$
$H \rightarrow \gamma\gamma$	$-5.8c'_\gamma$
$H \rightarrow \tau\tau$	$-1.0cH + 3.0cl$
$H \rightarrow gg$	$56c'_g$
$H \rightarrow \text{all}$	$0.0029cT + 0.17cu + 2.3cd + 0.11cl + 1.0cWW + 0.023cB + 0.37cHW$ $+0.0079cHB + 1.6c'_g + 0.0078cHQ + 0.17cpHQ + 0.0027cHu + 0.057cpHL$

	$\sum_{ij} B_{ij} c_i c_j$
$H \rightarrow b\bar{b}$	$0.25cH^2 + 2.3cd^2 + cH(-1.5cd)$
$H \rightarrow c\bar{c}$	$0.25cH^2 + 2.3cu^2 + cH(-1.5cu)$
$H \rightarrow \tau\tau$	$0.25cH^2 + 2.3cl^2 + cH(-1.5cl)$
$H \rightarrow \gamma\gamma$	$8.4(c_\gamma^2 + c_{\tilde{\gamma}}^2)$
$H \rightarrow gg$	$790(c_g^2 + c_{\tilde{g}}^2)$
$H \rightarrow WW^* \rightarrow l\nu l\nu$	$0.25cH^2 + 26cWW^2 + 3.8cHW^2 + 1.3cpHL^2 + 0.32tcHW^2$ $+cH(-5.1cWW - 1.9cHW - 1.1cpHL) + cWW(19cHW) + 12cpHL$ $+cHW(4.3cpHL)$
$H \rightarrow ZZ^* \rightarrow 4l$	$0.25cH^2 + 4.0cT^2 + 28cWW^2 + 3.5cB^2 + 2.2cHW^2 + 0.20cHB^2 + 1.8cHL^2$ $+1.8cpHL^2 + 0.43cHe^2 + 0.14tcHW^2 + cH(2.0cT - 5.1cWW - 1.3cB$ $-1.4cHW - 0.43cHB - 1.0cHL - 1.0cpHL + 0.43cHe) + cT(-21cWW$ $-5.3cB - 5.7cHW - 1.7cHB - 4.1cHL - 4.1cpHL + 1.7cHe) + cWW(10cB$ $+15cHW + 4.4cHB + 12cHL + 12cpHL - 3.5cHe) + cB(3.8cHW + 1.1cHB$ $+0.052c'_\gamma + 1.1cHL + 1.1cpHL - 2.1cHe) + cHW(1.3cHB + 3.0cHL$ $+3.0cpHL - 1.3cHe) + cHB(0.91cHL + 0.91cpHL - 0.39cHe)$ $+cHL(3.5cpHL - 0.13cHe) + cpHL(-0.13cHe) + 0.081tcHW(tcHB)$
$H \rightarrow \text{all}$	$0.24cH^2 + 0.037cT^2 + 0.13cu^2 + 1.7cd^2 + 0.084cl^2 + 2.6cWW^2 + 4.7cHW^2$ $+4.3cHB^2 + 23c'_g^2 + 0.09cpHQ^2 + 0.066cHu^2 + 0.027cpHL^2 + 4.3tcHW^2$ $+4.3tcHB^2 + 23c'_g^2 + cH(-0.086cu - 1.2cd - 0.056cl - 0.51cWW$ $-0.18cHW - 0.083cpHQ - 0.029cpHL) + cT(-0.19cWW - 0.046cB$ $-0.051cHW - 0.027cpHQ) + cWW(0.11cB + 1.9cHW + 0.04cHB + 0.86cpHQ$ $+0.29cpHL) + cHW(0.03cB - 8.6cHB + 0.1c_\gamma + 0.31cpHQ + 0.11cpHL)$ $+cHB(-0.1c_\gamma) + tcHW(-8.6tcHB + 0.1c'_{\tilde{\gamma}}) + tcHB(-0.10c'_{\tilde{\gamma}})$

Prepare measurements



You have 3 stage 0 categories and one decay:

ggH, qqH, ttH, x Hgg

```
OBJ: TH1F    data_ggH_hgg_mgg      Histogram of data_ggH_hgg_mgg : 0 at: 0x6adb920
OBJ: TH1F    data_qqH_hgg_mgg      Histogram of data_qqH_hgg_mgg : 0 at: 0x6c56870
OBJ: TH1F    data_ttH_hgg_mgg      Histogram of data_ttH_hgg_mgg : 0 at: 0x6da0750
OBJ: TH2D    Response      sig : 0 at: 0x6c602f0
```

The data are in the input file (Session 3/inputs_session3.root).

They are TH1Fs (you need to convert them into RooDataHists) eg.

In ROOT

```
TFile *fi = TFile::Open("inputs_session3.root");
RooRealVar x("x","x",110,160);
RooDataHist data1("data1","ggH_tagged_data",RooArgList(x),(TH1*)fi.Get("data_ggH_hgg_mgg"));
```

Or in PyROOT

```
import ROOT
fi = ROOT.TFile.Open("inputs_session3.root")
x = ROOT.RooRealVar("x","x",110,160)
data1 = ROOT.RooDataHist("data1","ggH_tagged_data",ROOT.RooArgList(x),fi.Get("data_ggH_hgg_mgg"))
```

Prepare measurements

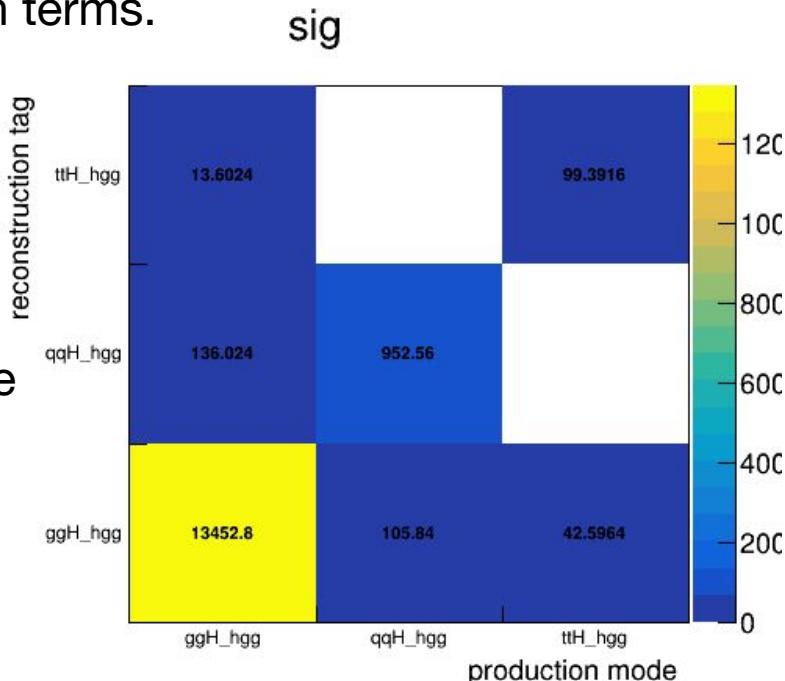


The expected signal events are written in the Response 2D Histogram with the relative confusion terms.

The signal is a gaussian with width $s = 1.5\text{GeV}$

The background is exponential

You can use what you learned in Session 1 to make the pdfs for the signal and background in each category.



Assignment: Part 1



Measure the Stage 0 cross section for ggH, qqH and ttH.

Use that the injected luminosity is $L=140 \text{ fb}^{-1}$

Derive the error and the covariance matrix of the fit.



Part 2

Measure directly the EFT params



If we float more than one EFT Parameter at the time, we need to check if we have a simultaneous sensitivity to all of them.

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Table 11: B_{jk} coefficients for the STXS stage 0 bins.

Measure directly the EFT params



Start from these scaling equations:

$$ggH : 1 + 8.73c'_G + 19.5(c'_G)^2$$

$$qqH : 1 + 0.6c_B - 0.797c_{HW} + 3.42c_B^2 + 114c_{HW}^2 + 15.3c_{HW}c_B$$

$$ttH : 1 + 0.115c'_G + c_B^2 + 5.87c_{HW}^2 + 0.0297(c'_G)^2 - 0.318c_Bc_{HW}$$

And derive the equivalent functions in terms of c'_G and $c_{HW}-c_B$ (assuming $c_{HW}+c_B=0$)

Now remake your signal models such that the scaling functions for the ggH, qqH and ttH processes are the ones you just derived

*Note that we have included the large factors of 10 inside the coefficient c_G and called it c'_G simply to avoid fitting with very small numbers



Part 3

Use the covariance matrix and the measurements you obtained in part 1 to derive the constraints on the parameter c_G .

Limitations:

- Gaussian approximation
- Nuisance parameters and correlation among them are (partially) neglected accordingly to what available.

Conclusions

Summary & Conclusions



We covered ...

- Some basic RooFit
 - Object creation/manipulation, pdfs and toy-generation, likelihood construction and minimization
- Simultaneous likelihoods
 - Multiple bins / multiple categories of data → additional constraints on physics parameters from including additional data in the LH
- Physics parameter determination
 - Multiple signal processes can contribute to multiple categories (regions) in data → unfolding allows to extract those contributions (cross-sections)
 - Being able to determine differences from expected contributions can be used to constraint EFT coefficients from the experimental data



Thanks!

