



Fitting: Session 3

Andrea C. Marini, Nick Wardle

Assistants:

Lydia Brenner

Sebastian Wuchterl

Adinda de Wit





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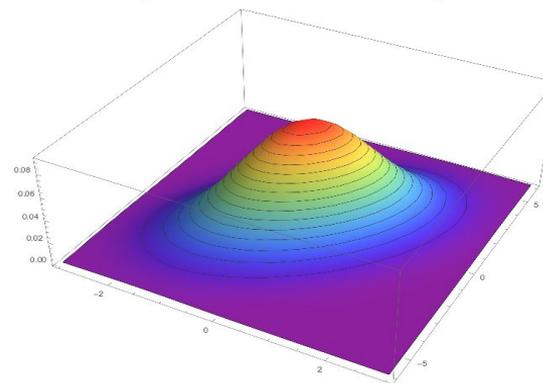
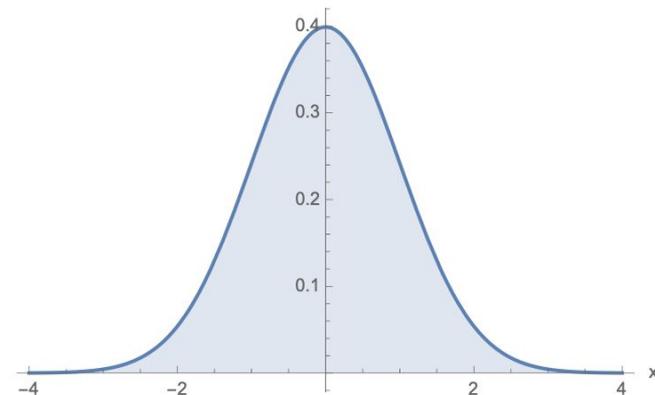
Statistical introduction

Continuous 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

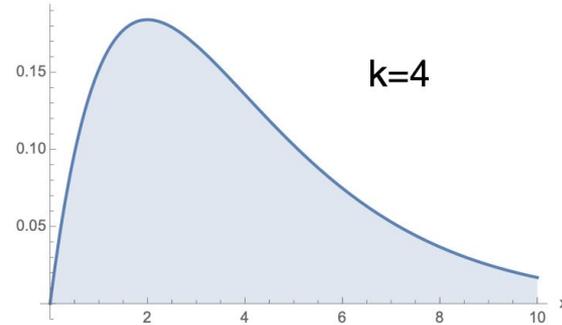
$$X_i \sim \mathcal{N}(0, 1)$$

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

k is the number
of degrees of freedom

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

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



Wilks' theorem



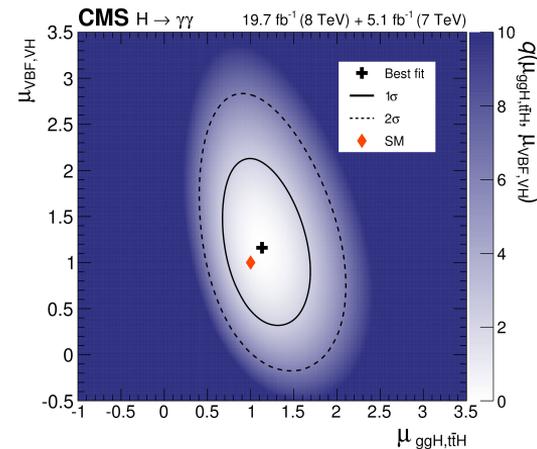
Likelihood ratio are often used in statistical tests:

- 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)

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

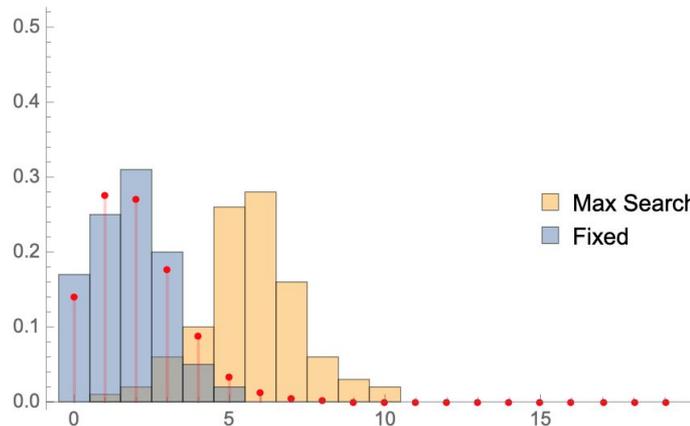
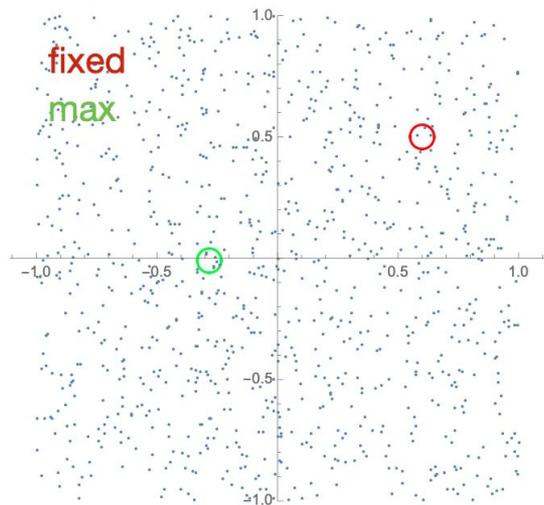
	68.3%	95%
1 ndf	1	3.84
2 ndf	2.23	5.99



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/qq \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/qq \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	$\sum_i A_i c_i$
$H \rightarrow b\bar{b}$	$-1.0c_H + 3.0c_d$
$H \rightarrow WW^* \rightarrow l\nu l\nu$	$10c_{WW} + 3.7c_{HW} + 2.2c_{PHL}$
$H \rightarrow ZZ^* \rightarrow 4l$	$55c_{WW} + 13c_B + 15c_{HW} + 4.6c_{HB} + 0.018c_\gamma + 2.0c_{HL} + 2.0c_{PHL} + 0.027c_{He}$
$H \rightarrow \gamma\gamma$	$-5.8c'_\gamma$
$H \rightarrow \tau\tau$	$-1.0c_H + 3.0c_l$
$H \rightarrow gg$	$56c'_g$
$H \rightarrow \text{all}$	$0.0029c_T + 0.17c_u + 2.3c_d + 0.11c_l + 1.0c_{WW} + 0.023c_B + 0.37c_{HW} + 0.0079c_{HB} + 1.6c'_g + 0.0078c_{HQ} + 0.17c_{PHQ} + 0.0027c_{Hu} + 0.057c_{PHL}$

Partial width	$\sum_{ij} B_{ij} c_i c_j$
$H \rightarrow b\bar{b}$	$0.25c_H^2 + 2.3c_d^2 + c_H(-1.5c_d)$
$H \rightarrow c\bar{c}$	$0.25c_H^2 + 2.3c_u^2 + c_H(-1.5c_u)$
$H \rightarrow \tau\tau$	$0.25c_H^2 + 2.3c_l^2 + c_H(-1.5c_l)$
$H \rightarrow \gamma\gamma$	$8.4(c'_\gamma)^2 + c'_\gamma c'_\gamma$
$H \rightarrow gg$	$790(c'_g)^2 + c'_g c'_g$
$H \rightarrow WW^* \rightarrow l\nu l\nu$	$0.25c_H^2 + 26c_{WW}^2 + 3.8c_{HW}^2 + 1.3c_{PHL}^2 + 0.32c_{HW}^2 + c_H(-5.1c_{WW} - 1.9c_{HW} - 1.1c_{PHL}) + c_{WW}(19c_{HW}) + 12c_{PHL} + c_{HW}(4.3c_{PHL})$
$H \rightarrow ZZ^* \rightarrow 4l$	$0.25c_H^2 + 4.0c_T^2 + 28c_{WW}^2 + 3.5c_B^2 + 2.2c_{HW}^2 + 0.20c_{HB}^2 + 1.8c_{HL}^2 + 1.8c_{PHL}^2 + 0.43c_{He}^2 + 0.14c_{HW}^2 + c_H(2.0c_T - 5.1c_{WW} - 1.3c_B - 1.4c_{HW} - 0.43c_{HB} - 1.0c_{HL} - 1.0c_{PHL} + 0.43c_{He}) + c_T(-21c_{WW} - 5.3c_B - 5.7c_{HW} - 1.7c_{HB} - 4.1c_{HL} - 4.1c_{PHL} + 1.7c_{He}) + c_{WW}(10c_B + 15c_{HW} + 4.4c_{HB} + 12c_{HL} + 12c_{PHL} - 3.5c_{He}) + c_B(3.8c_{HW} + 1.1c_{HB} + 0.052c'_\gamma + 1.1c_{HL} + 1.1c_{PHL} - 2.1c_{He}) + c_{HW}(1.3c_{HB} + 3.0c_{HL} + 3.0c_{PHL} - 1.3c_{He}) + c_{HB}(0.91c_{HL} + 0.91c_{PHL} - 0.39c_{He}) + c_{HL}(3.5c_{PHL} - 0.13c_{He}) + c_{PHL}(-0.13c_{He}) + 0.081c_{HW}(c_{HB})$
$H \rightarrow \text{all}$	$0.24c_H^2 + 0.037c_T^2 + 0.13c_u^2 + 1.7c_d^2 + 0.084c_l^2 + 2.6c_{WW}^2 + 4.7c_{HW}^2 + 4.3c_{HB}^2 + 23c'_g{}^2 + 0.09c_{PHQ}^2 + 0.066c_{Hud}^2 + 0.027c_{PHL}^2 + 4.3c_{HW}^2 + 4.3c_{HB}^2 + 23c'_g{}^2 + c_H(-0.086c_u - 1.2c_d - 0.056c_l - 0.51c_{WW} - 0.18c_{HW} - 0.083c_{PHQ} - 0.029c_{PHL}) + c_T(-0.19c_{WW} - 0.046c_B - 0.051c_{HW} - 0.027c_{PHQ}) + c_{WW}(0.11c_B + 1.9c_{HW} + 0.04c_{HB} + 0.86c_{PHQ} + 0.29c_{PHL}) + c_{HW}(0.03c_B - 8.6c_{HB} + 0.1c_\gamma + 0.31c_{PHQ} + 0.11c_{PHL}) + c_{HB}(-0.1c_\gamma) + c_{HW}(-8.6c_{HB} + 0.1c'_\gamma) + c_{HB}(-0.10c'_\gamma)$

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

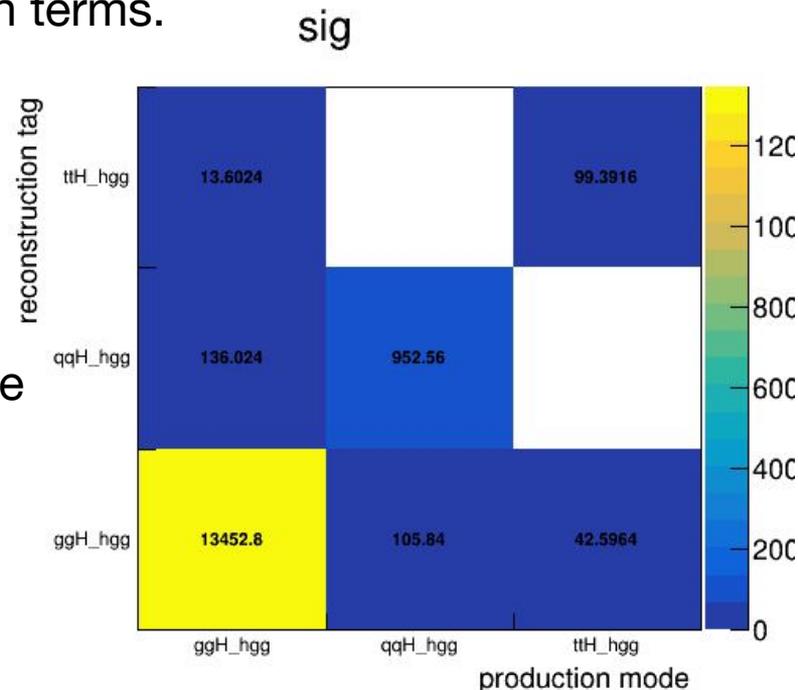


The expected signal events are written in the Response 2D Histogram with the relative confusion terms.

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.

The production scaling depends on c_G , c_{WW} , c_B , c_{HW} , c_A , ...

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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

Prop. the Cov matrix to the EFT



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!

