

# Basics in statistical analysis and hypothesis testing for physicists: Exercises

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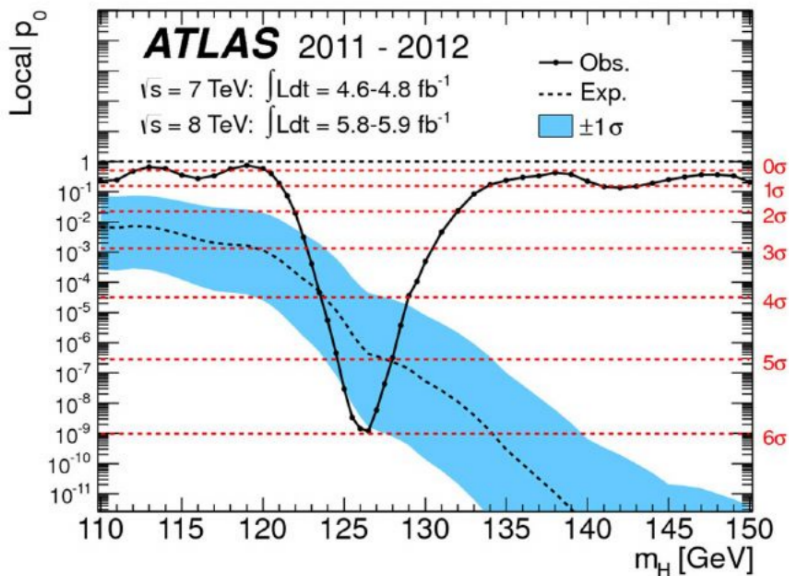
**Gluo**dynamic*s*



# Outline

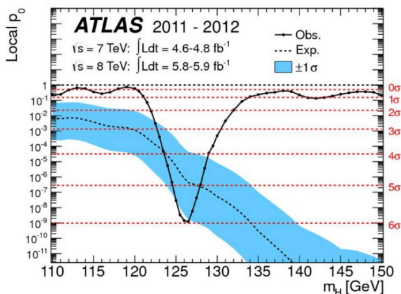
- ▶ Warm-up: reading a discovery plot
- ▶ Ex 1: A published  $\chi^2$ -fit
- ▶ Ex 2: Two cross sections and its ratio
- ▶ Ex 3: Simple exercise: How good is my model?
- ▶ Ex 4: Simple hypothesis test

# Warm-up



# Warm-up

- Each Higgs mass hypothesis is scanned independently
- For each mass:
  - **Observed**: p-value observed in data
  - **Expected**: median of the p-value expected in the presence of the SM Higgs boson
  - Blue band: interval containing 68% of the p-values under SM Higgs hypothesis
- “Local”  $p_0$ 
  - Many mass points scanned
  - **Look-elsewhere effect**: global  $p_0$  to correct for number of trials



## Ex 1: A published $\chi^2$ -fit

- ▶ a least  $\chi^2$ -fit
- ▶ the reported  $\chi^2/ndf$  is 1.72
- ▶ the distribution is a mass distribution containing known resonance peaks, no weights are used
- ▶ the content of each bin follows a Poisson distribution
- ▶ the data satisfy well the Gaussian approximation for a Poisson since in each bin  $N_{yield,bin} \gg 10$
- ▶ Is this a good fit based on the chi2-test?

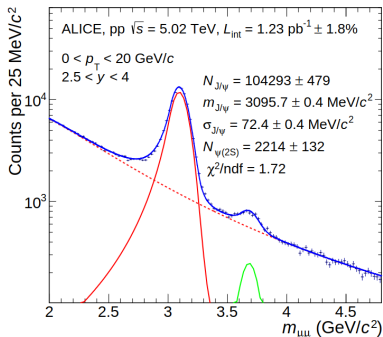
# Ex 1

- ▶ we can't tell without knowing the degrees of freedom (NDF)!
- ▶ NDF about 110
- ▶ Question: is this a good fit?

# Ex 1

- ▶ for  $N > 100$ ,  $\chi^2$ -distribution approximately gaussian with mean  $NDF$  and variance  $2 \cdot NDF$
- ▶ to reject the  $H_0$ , we choose a 5-sigma deviation
- ▶ sigma of Gaussian is 14.8
- ▶ we can exclude fit model at more than  $5 \sigma$   
→ a new resonance in the fit, a discovery?

# Ex 1



[arXiv:2109.15240](https://arxiv.org/abs/2109.15240)

- ▶ Reasons for the model doing so bad?
- ▶ Why accepted by collaboration/peer-review?



# Ex 1

- ▶ Reasons: peak-parameters vary as function of integrated kinematics, would need more complicated model
- ▶ signal extraction one contributor to uncertainties, differential results available  
→ certainly not ideal

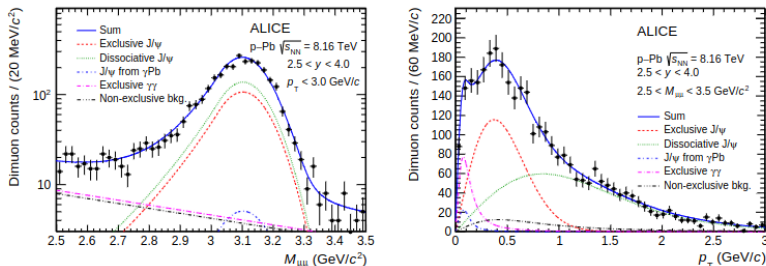
## Advice:

- ▶ think about it through before what makes sense
- ▶ visualise the data
- ▶ Always quote NDF if you do chi2-fits
- ▶ if the p-value of a test is very bad, but makes no sense due to systematic effects not taking into account of the model → identify the systematic effects, figure out their importance, correct them if possible

## Ex2: Two cross sections and its ratio

- ▶ What is your expectation for the statistical uncertainty on a ratio vs. the uncertainties between numerator and denominator?  
→ how would you guess the statistical uncertainty of the ratio between the two cross sections?

## Ex2: Two cross sections and its ratio



**Figure 2:** Projections of the two-dimensional fit on the dimuon invariant mass (left) and  $p_T$  (right).

- ▶ 2D unbinned maximum-loglikelihood fit, both cross sections from peak in the left, different shape on the right
- ▶ what does this tell you about the statistical uncertainty of the cross section ratio?

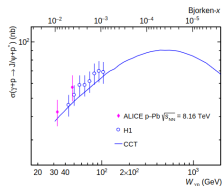
## Ex2: Two cross sections and its ratio

**Table 4:** Ratio of dissociative-to-exclusive  $J/\psi$  photoproduction cross sections in p-Pb UPCs at  $\sqrt{s_{NN}} = 8.16$  TeV. The first uncertainty is the statistical one. Its size is strongly impacted by the anti-correlation between exclusive and dissociative  $J/\psi$  components in the 2-dimensional fit. The second uncertainty is the systematic one. It is computed as the quadratic sum of the signal extraction ratio uncertainty, and the uncertainty on the V0C veto.

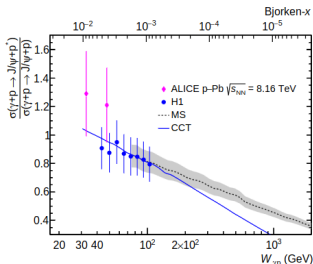
Rapidity range	$W_{\gamma p}$ (GeV)	$\langle W_{\gamma p} \rangle$ (GeV)	$\frac{\sigma(\gamma + p \rightarrow J/\psi + p^{(*)})}{\sigma(\gamma + p \rightarrow J/\psi + p)}$
(2.5, 4)	(27, 57)	39.9	$1.27 \pm 0.15 \pm 0.18$
(3.25, 4)	(27, 39)	32.8	$1.29 \pm 0.23 \pm 0.19$
(2.5, 3.25)	(39, 57)	47.7	$1.21 \pm 0.18 \pm 0.18$

- ▶ anticorrelation, e.g. for integrated cross section result:  
linear error propagation assuming independence would give 8.8%  
in reality considering anticorrelation 11.8%

## Ex2: Two cross sections and its ratio



**Figure 6:** Dissociative  $J/\psi$  photoproduction cross section off protons measured by ALICE in p-Pb UPCs at  $\sqrt{s_{NN}} = 8.16$  TeV and compared with H1 data [40]. A comparison with the CCT model [37] is shown. The uncertainties of the data points are the quadratic sum of the statistical and systematic uncertainties.

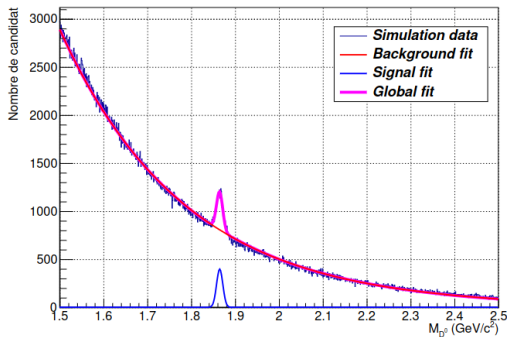


- ▶ can be very different in different experiments

### Advice

- ▶ Provide correlation matrices for uncertainties if possible
- ▶ be careful with assumption transfer from one situation to another

## Ex3: Simple exercise



Result of a fast simulation

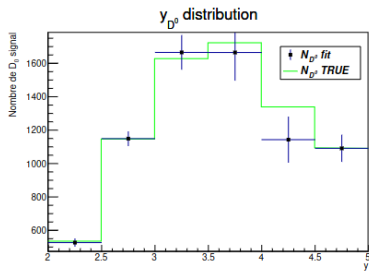
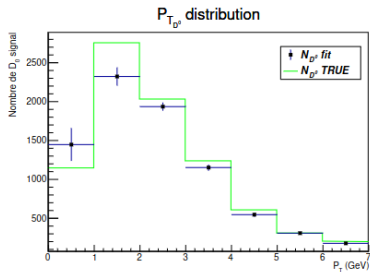
simple chi2-fit with exponential background, Gaussian signal; good fit quality

Side question: reasonable range and binning choice?

simulation: we know the truth!

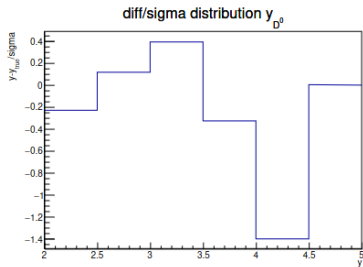
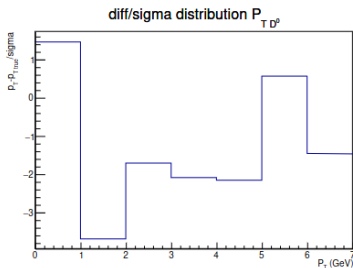
let's check

## Ex3: Simple exercise



► is this worrisome?

## Ex3 Simple exercise



- ▶ is this worrisome?
- ▶ what can we do to check:
  - 1) bias
  - 2) convergence
  - 3) correct coverage of fit?



## Ex4: Simple example

Counting experiment for so far unobserved signal:

- ▶  $D$  observed events: measurement outcome
- ▶  $s$  expected signal events: 7.3
- ▶  $b$  expected background events (known): 2.6

What is the correct model PDF?

## Ex4: model PDF

A Poissonian distribution:

$$p(D|C) = \frac{(C)^D e^{-C}}{D!}$$

Consider  $H_0$ : background only,  $C = B_0$ ;  $H_1$ : background and signal  $C = S + B_0$

## Ex4: Parameter estimation

Use Maximum Likelihood to estimate signal counts provide  $b$  known and  $D$  observed events

$$L(S) = p(D|S, B_0)$$

$$dL/dS = 0 \Rightarrow S = D - B_0$$

## Ex4: Simple example discovery

do we expect a  $5\sigma$  discovery or a  $3\sigma$  *evidence*?

when do we have a  $5\sigma$  discovery?

$H_0$ : background only,  $C = B_0$ ,  $H_1$ : signal according to simulation,  $C = S + B_0$

Make simple hypothesis testing, without unspecified parameters,

see [arXiv:1007.1727v3](https://arxiv.org/abs/1007.1727v3) for composite

Use the maximal likelihood ratio as test statistics due to Neyman-Pearson lemma

## Ex4: The test statistics, likelihood ratio

Our data is one number:  $D$

The parameter of interest, model expectation  $\Theta = C$ :

$C = B_0(H_0)$ ,  $C = S + B_0(H_1)$

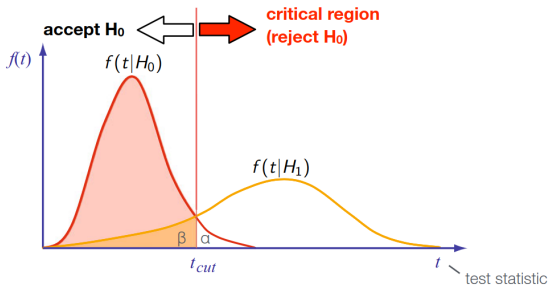
$$t(D) = \frac{L(D|H_1)}{L(D|H_0)} \quad (1)$$

$$= \frac{(S+B)^D e^{-(S+B)} D!}{D! B^D e^{-B}} \quad (2)$$

$$\propto \left( \frac{S+B_0}{B_0} \right)^D \quad (3)$$

significance of test:  $2.9 \cdot 10^{-7}$  ( $5\sigma$ ), what do we need to know determine?

## Ex4: critical region



The probability for  $H_0$  to be rejected while  $H_0$  is true:

$$\int_{t_{cut}}^{\infty} f(t|H_0) dt = \alpha$$

$\alpha$ :  
"size" or "significance level" of the test

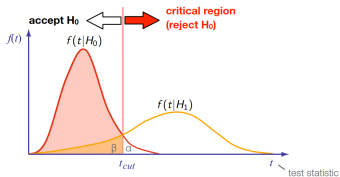
Probability to reject  $H_1$  even though it is true:

$$\int_{-\infty}^{t_{cut}} f(t|H_1) dt = \beta$$

$1 - \beta$ :  
"power of the test",  
prob. to reject  $H_0$  if  $H_1$  is true

we need  $f(t|H_1)$  and  $f(t|H_0)$

## Ex4: critical region



The probability for  $H_0$  to be rejected while  $H_0$  is true:

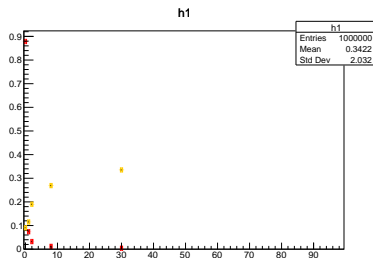
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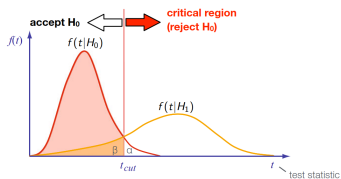


Example: light-by-light scattering, [DOI:10.1038/NPHYS4208](https://doi.org/10.1038/NPHYS4208)

$B_0 = 2.6$ ,  $S = 7.3$  (expectation, neglect uncertainties)

$\ln \frac{L(D|H_1)}{L(D|H_0)} = D \cdot \text{const}$ : Take rather  $D$  as a  $t$

## Ex4: critical region



The probability for  $H_0$  to be rejected while  $H_0$  is true:

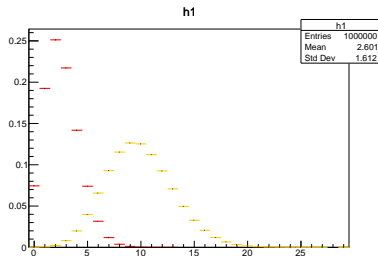
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 $B_0 = 2.6$ ,  $S = 7.3$  (expectation, neglect uncertainties)

$\rightarrow t = D$

$t_{cut}$  for 5  $\sigma$ :  $D = 16$

$t_{cut}$  for 3  $\sigma$ :  $D = 9$

$\rightarrow$  we expect evidence, but no discovery



## Ex4: Simple hypothesis testing

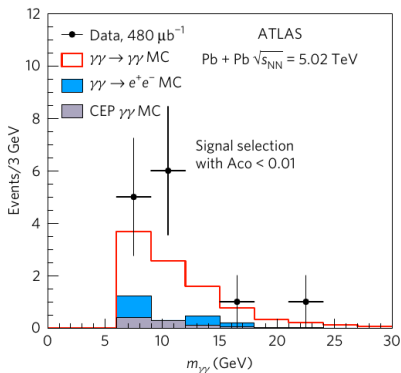
Example: light-by-light scattering evidence,  
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$B_0 = 2.6$ ,  $S = 7.3$  (expectation)

Observation:  $D = 13$ :

evidence, but no discovery with our t

## Ex4: Simple hypothesis testing



Example: light-by-light scattering evidence, [DOI:10.1038/NPHYS4208](https://doi.org/10.1038/NPHYS4208)

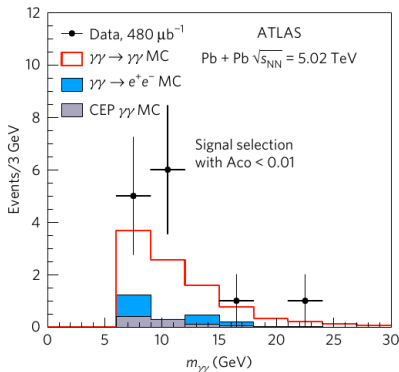
$B_0 = 2.6$ ,  $S = 7.3$  (expectation)

Observation:  $D = 13$

p-value with  $t = D$  neglecting uncertainties:

$\approx 2 \cdot 10^5$

## Ex4: example



To quantify an excess of events over the background expectation, a test statistic based on the profile likelihood ratio<sup>18</sup> is used. The  $p$  value for the background-only hypothesis, defined as the probability for the background to fluctuate and give an excess of events as large or larger than that observed in the data, is found to be  $5 \times 10^{-6}$ . The  $p$  value can be expressed in terms of Gaussian tail probabilities, which, given in units of standard deviation ( $\sigma$ ), corresponds to a significance of  $4.4\sigma$ . The expected  $p$  value and significance (obtained before the fit of the signal-plus-background hypothesis to the data and using standard model predictions from ref. 28) are  $8 \times 10^{-5}$  and  $3.8\sigma$ , respectively.