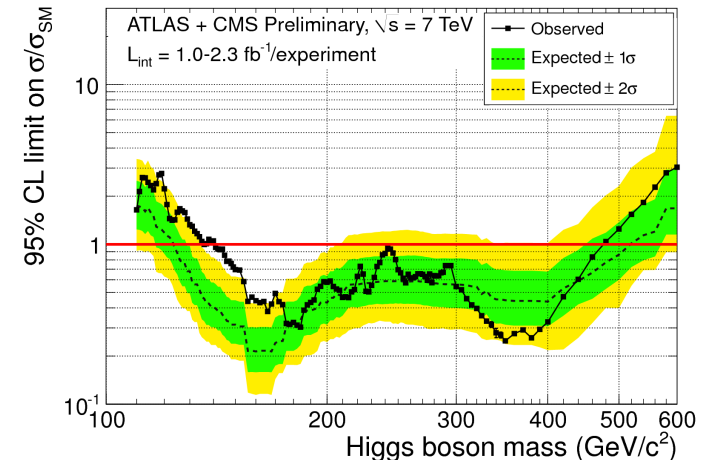
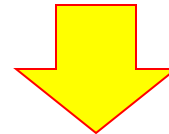
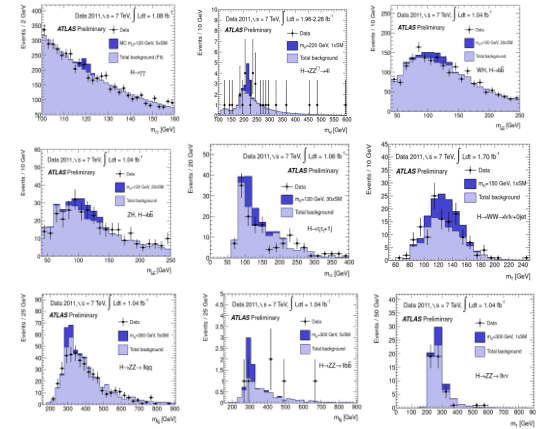


Statistical aspects of Higgs analyses

W. Verkerke
(NIKHEF)

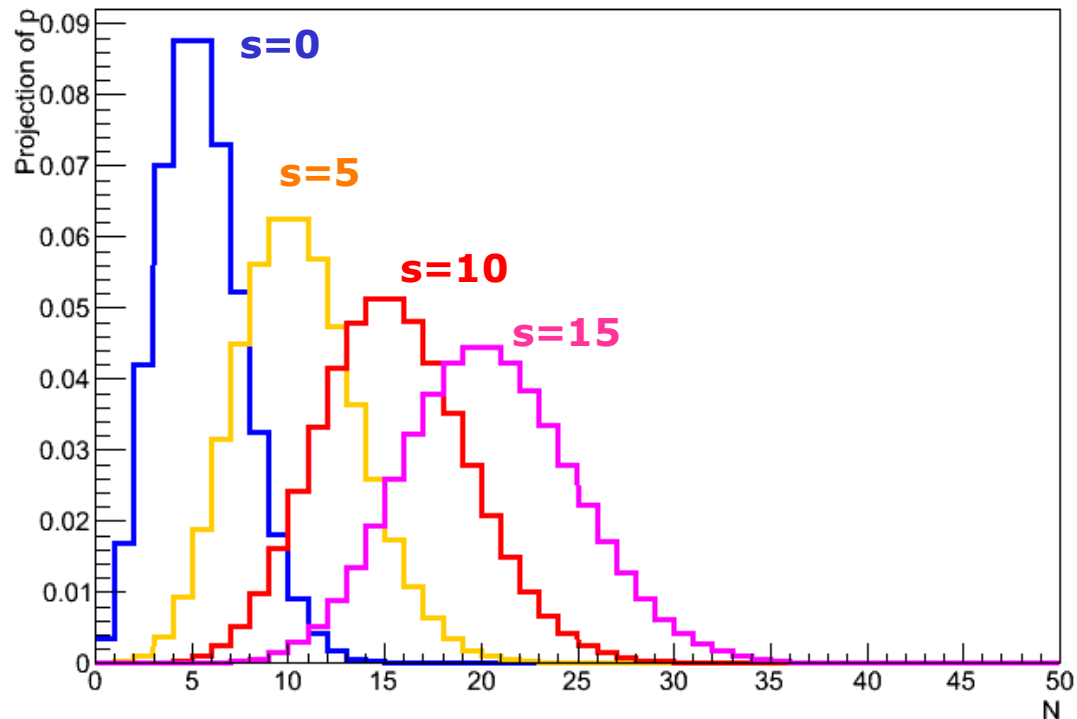
Introduction

- Enormous effort to search for Higgs signature in many decay channels
- Results \rightarrow many plots with signal, background expectations, each with (systematic) uncertainties, and data
- Q: How do you conclude from this that you've seen the Higgs (or not)?
 - Want answer of type: "We can exclude that the Higgs exist at 95% CL", or "Probability that background only caused observed excess is $3 \cdot 10^{-7}$ "
- Here a short guide through how this is (typically) done



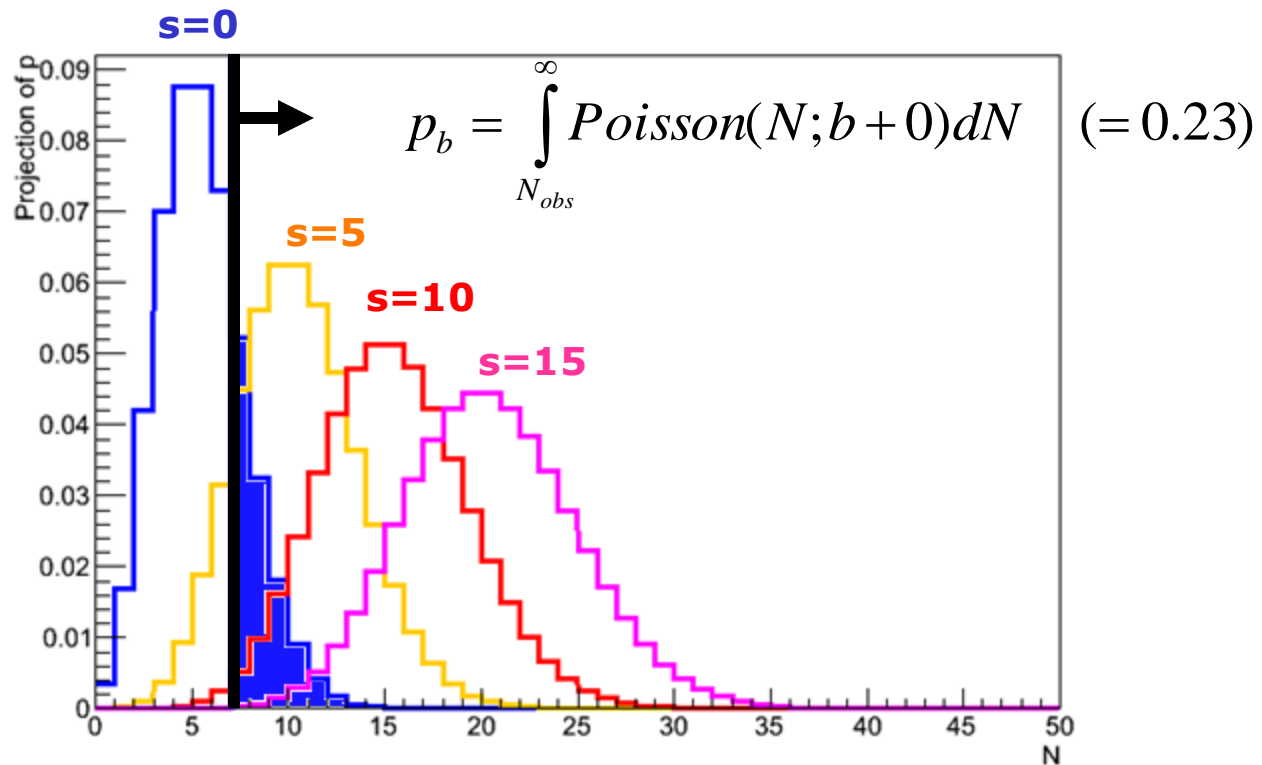
Quantifying discovery and exclusion – Frequentist approach

- Consider the simplest case – a counting experiment
 - Observable: \mathbf{N} (the event count)
 - Model $F(\mathbf{N}|\mathbf{s})$: Poisson($\mathbf{N}|\mathbf{s}+b$) with $b=5$ known exactly
- Predicted distributions of \mathbf{N} for various values of \mathbf{s}



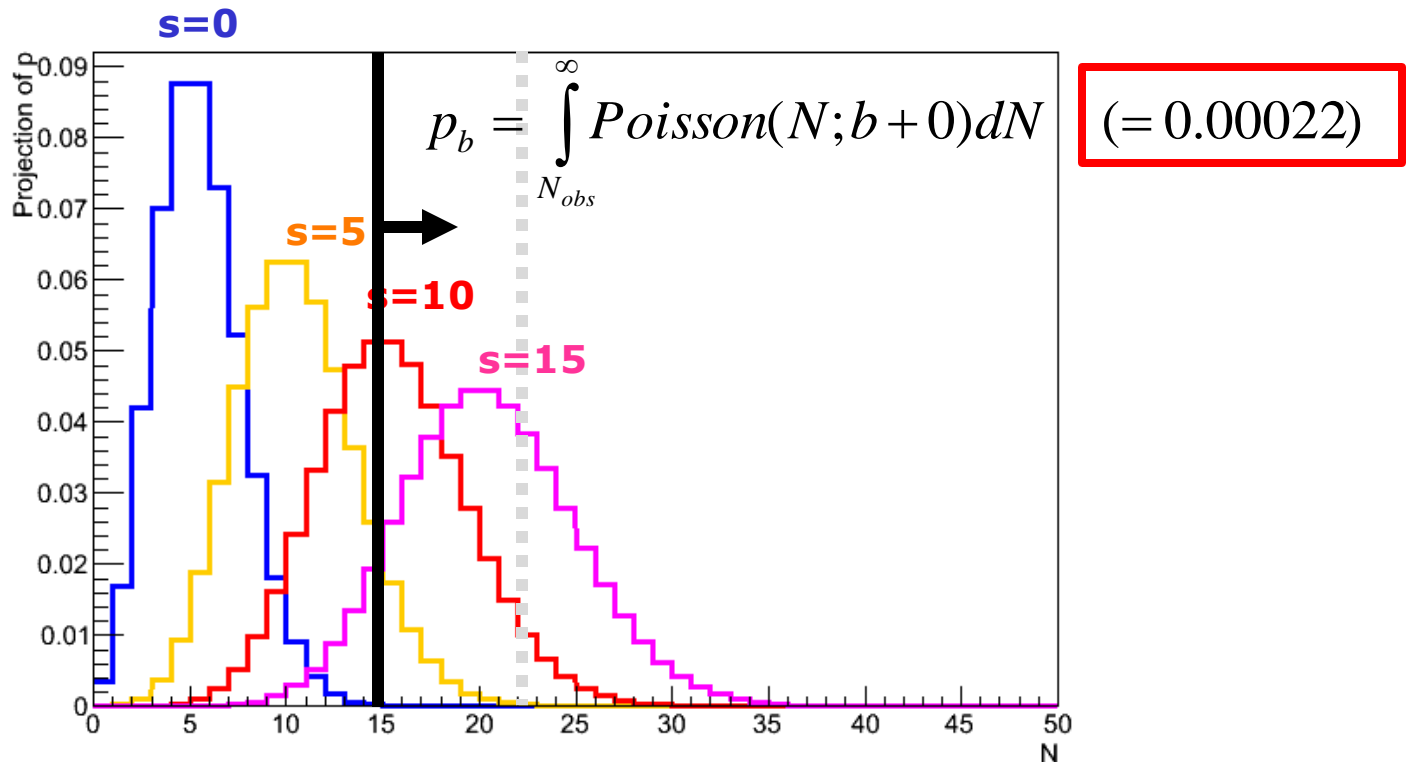
Frequentist p-values – excess over expected bkg

- Now make a measurement $N=N_{\text{obs}}$ (example $N_{\text{obs}}=7$)
- Can now define p-value(s), e.g. for bkg hypothesis
 - Fraction of future measurements with $N=N_{\text{obs}}$ (or larger) if $s=0$



Frequentist p-values - excess over expected bkg

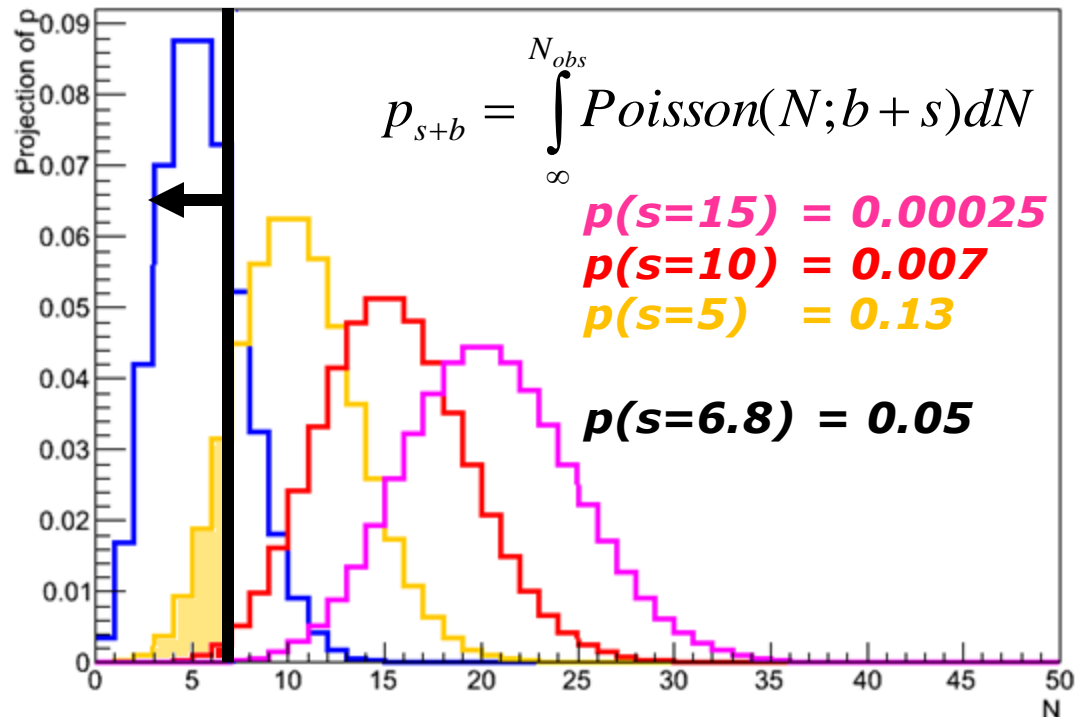
- p-values of background hypothesis is used to quantify 'discovery' = excess of events over background expectation
- Another example: $N_{obs}=15$ for same model, what is p_b ?



- Result customarily re-expressed as odds of a *Gaussian fluctuation with equal p-value* (3.5 sigma for above case)
- NB: $N_{obs}=22$ gives $p_b < 2.8 \cdot 10^{-7}$ ('5 sigma')

Upper limits (one-sided confidence intervals)

- Can also defined p-values for **hypothesis with signal**: p_{s+b}
 - Note convention: integration range in p_{s+b} is flipped



- Convention: express result as value of **s** for which $p(s+b)=5\%$ → **“s>6.8 is excluded at 95% C.L.”**

Modified frequentist upper limits

- Need to be careful about interpretation $p(s+b)$ in terms of inference on signal only
 - Since $p(s+b)$ quantifies consistency of signal *plus* background
 - Problem most apparent when observed data has **downward stat. fluctuations w.r.t background** expectation

- Example: $N_{\text{obs}} = 2$

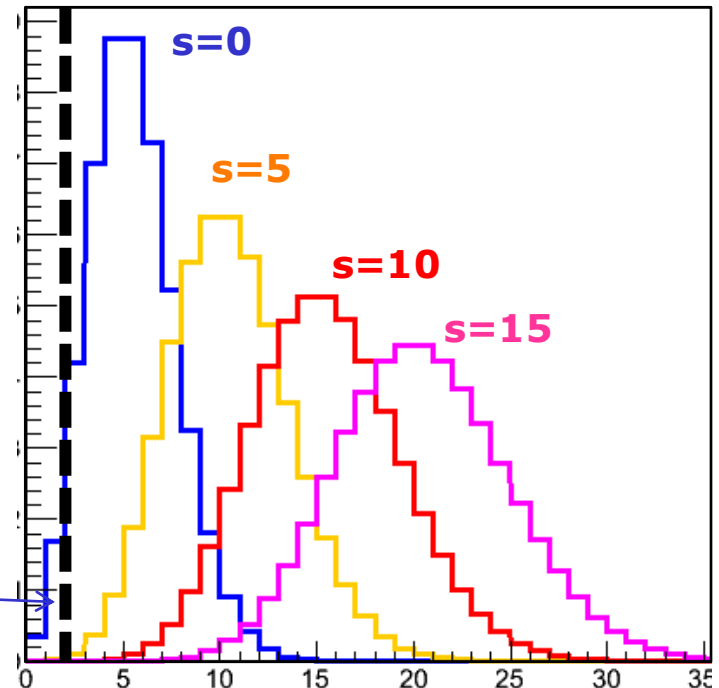
$$\rightarrow p_{s+b}(s=0) = 0.04$$

$s \geq 0$ excluded at >95% C.L. ?!

- **Modified approach to protect against such inference on s**

- Instead of requiring $p(s+b) = 5\%$, require

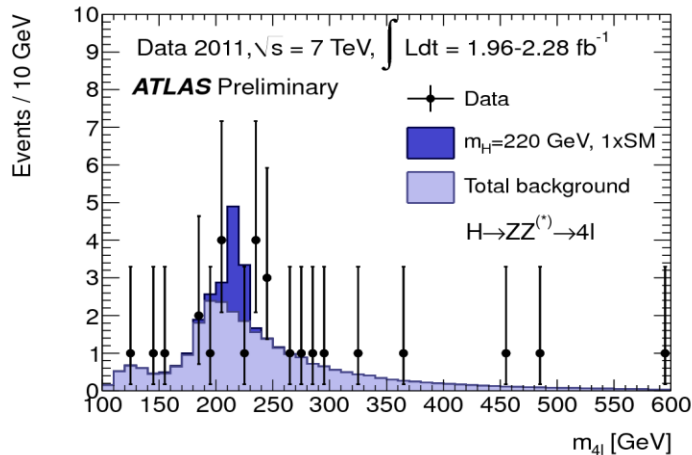
$$CL_s \equiv \frac{P_{s+b}}{1 - p_b} = 5\%$$



for $N=2$ exclude $s > 3.4$ at 95% C.L.s, for large N effect on limit is small as $p_b \rightarrow 0$

p-values and limits on non-trivial analysis

- Typical Higgs search result is not a simple number counting experiment, but looks like this:



- Result is a distribution, not a single number
- Models for signal and background have intrinsic uncertainties

- Any type of result can be converted into a single number by constructing a 'test statistic'
 - **A test statistic compresses all signal-to-background discrimination power in a single number**
 - Most powerful discriminators are **Likelihood Ratios** (Neyman Pearson)

$$q_{\mu} = -2 \ln \frac{L(\text{data} | \mu)}{L(\text{data} | \hat{\mu})}$$

The likelihood ratio test statistic

- Definition: $\mu = \text{signal strength} / \text{signal strength(SM)}$
 - Choose e.g likelihood with nominal signal strength in numerator ($\mu=1$)

'likelihood assuming nominal signal strength'

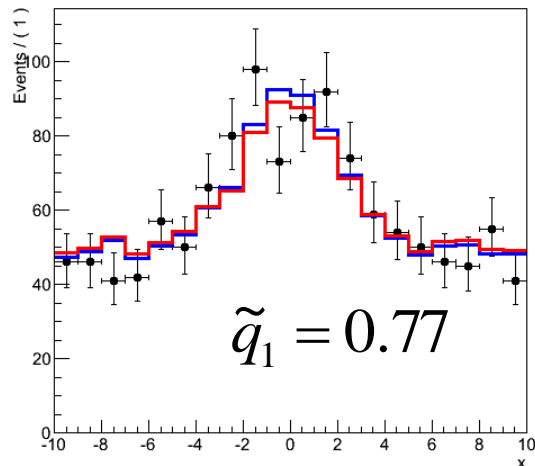
$$q_1 = -2 \ln \frac{L(\text{data} | \mu = 1)}{L(\text{data} | \hat{\mu})}$$

$\hat{\mu}$ is best fit value of μ

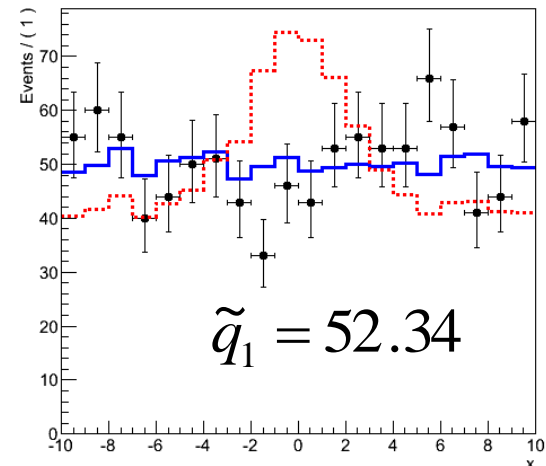
'likelihood of best fit'

- Illustration on model with no shape uncertainties

On signal-like data q_1 is **small**

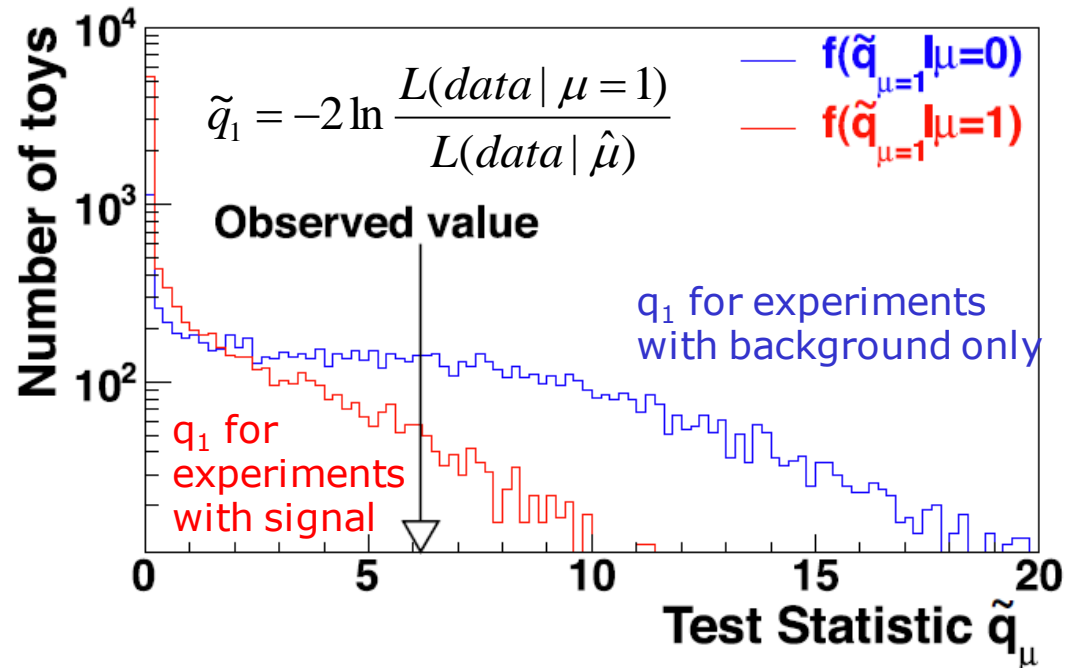
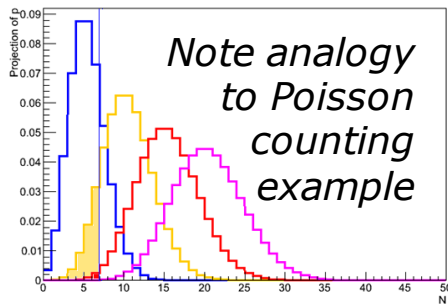
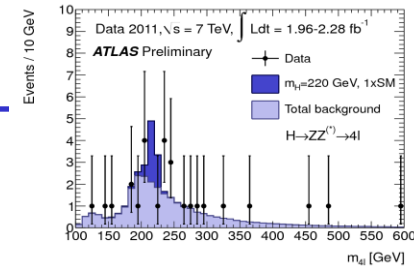


On background-like data q_1 is **large**



Distributions of test statistics

- Value of q_1 on data is now the 'measurement'
- Distribution of q_1 *not* calculable
 - But can obtain distribution from pseudo-experiments
 - Generate a large number of pseudo-experiments with a given value of μ , calculate q for each, plot distribution



- From q_{obs} and these distributions, can then set limits similar to what was shown for Poisson counting example

Incorporating systematic uncertainties

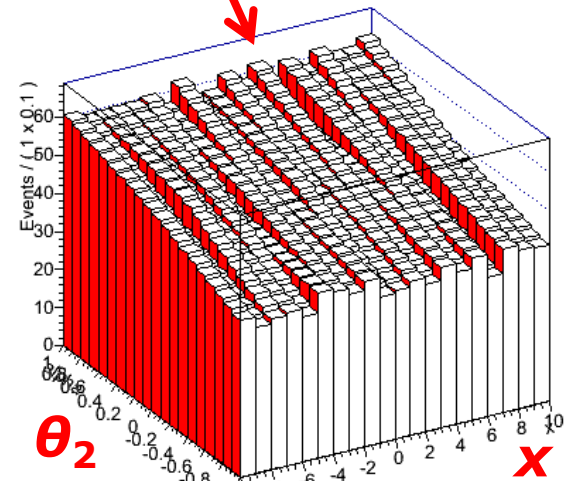
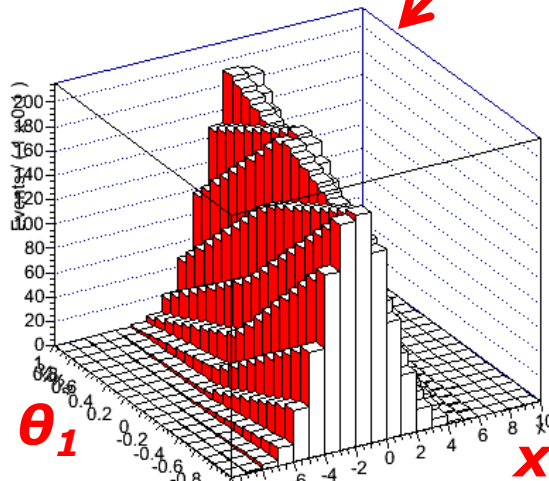
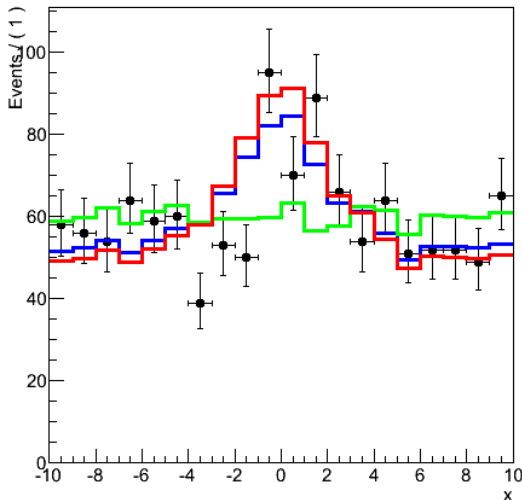
- What happens if models have uncertainties

$$L(\text{data} | \mu) \rightarrow L(\text{data} | \mu, \vec{\theta})$$

Jet Energy Scale,
QCD scale,
luminosity,
...

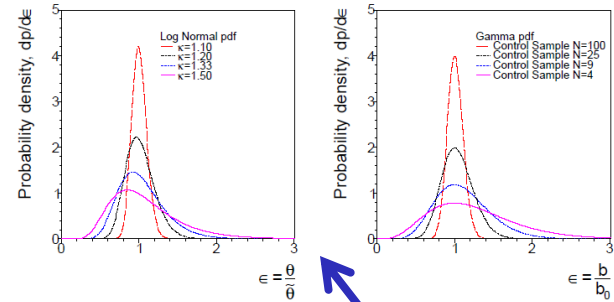
- Introduction of additional model parameters θ that describe effect of uncertainties

$$L(\text{data} | \mu, \theta) = \text{Poisson}(N_i | \mu \cdot s_i(\theta) + b_i(\theta)) \cdot p(\vec{\theta}, \theta)$$

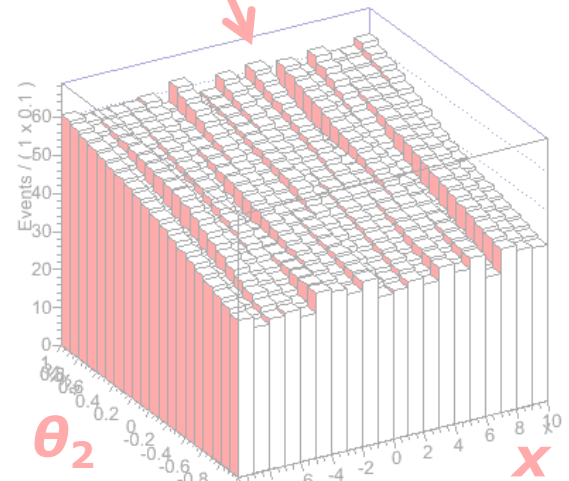
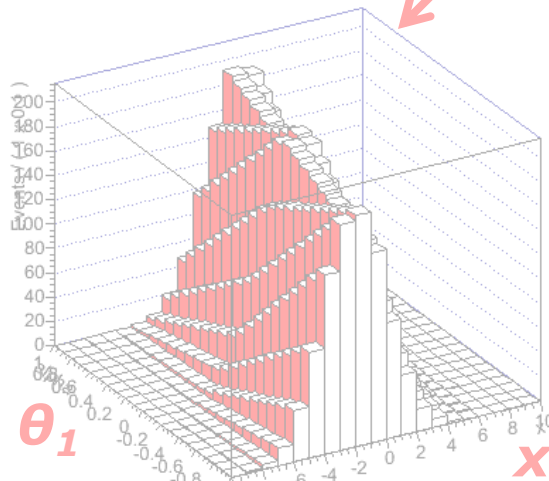
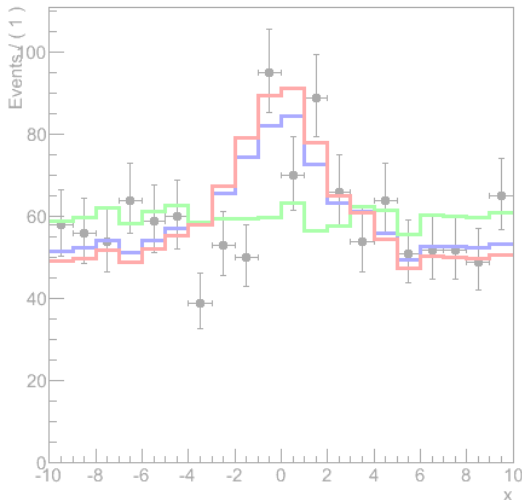


Incorporating systematic uncertainties

Likelihood includes *auxiliary measurement* terms that constrains the nuisance parameters θ (shape is flat, log-normal, gamma, or Gaussian)




$$L(\text{data} | \mu, \theta) = \text{Poisson}(N_i | \mu \cdot s_i(\theta) + b_i(\theta)) \cdot p(\tilde{\theta}, \theta)$$



Dealing with nuisance parameters in the test statistic

- Uncertainty quantified by nuisance parameters are incorporated in test statistic using a **profile likelihood ratio**

$$q_\mu = -2 \ln \frac{L(\text{data} | \mu)}{L(\text{data} | \hat{\mu})}$$

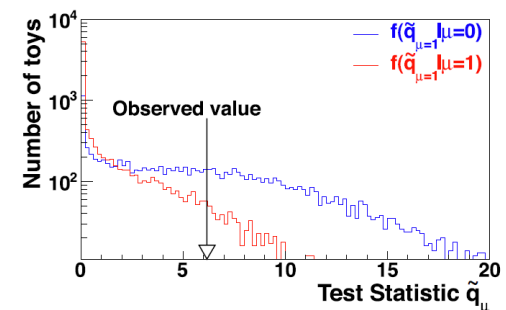

$$\tilde{q}_\mu = -2 \ln \frac{L(\text{data} | \mu, \hat{\theta}_\mu)}{L(\text{data} | \hat{\mu}, \hat{\theta})}$$

'likelihood of best fit for a given fixed value of μ '

'likelihood of best fit'

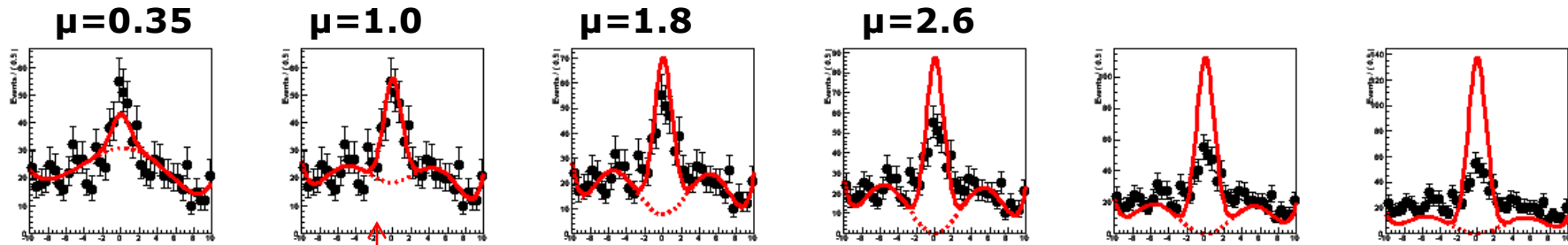
(with a constraint $0 \leq \hat{\mu} \leq \mu$)

- Interpretation of *observed* value of q_μ in terms of p-value again based on expected distribution obtained from pseudo-experiments



Dealing with nuisance parameters in the test statistic

- Uncertainty quantified by nuisance parameters are incorporated using a **profile likelihood ratio test statistic**



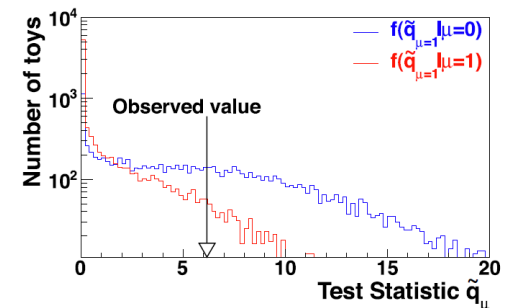
$$\tilde{q}_\mu = -2 \ln \frac{L(\text{data} | \mu, \hat{\theta}_\mu)}{L(\text{data} | \hat{\mu}, \hat{\theta})}$$

(with a constraint $0 \leq \hat{\mu} \leq \mu$)

'likelihood of best fit for a given fixed value of μ '

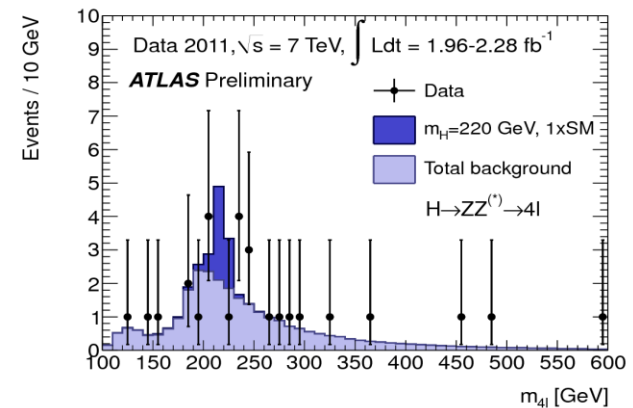
'likelihood of best fit'

- Interpretation of *observed* value of q_μ in terms of p-value again based on expected distribution obtained from pseudo-experiments



Putting it all together for one Higgs channel

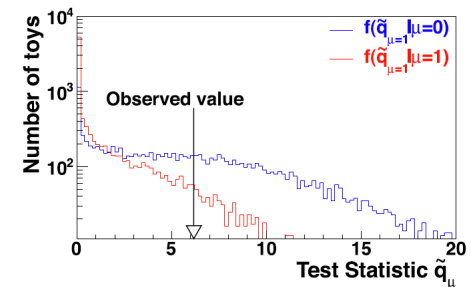
- Result from data $D(x)$
- PDF $F(x | m_H, \mu, \theta)$ that models the data for a given true Higgs mass hypothesis



$$\tilde{q}_\mu(m_H) = -2 \ln \frac{L(\text{data} | \mu, m_H \hat{\theta}_\mu)}{L(\text{data} | \hat{\mu}, m_H \hat{\theta})}$$

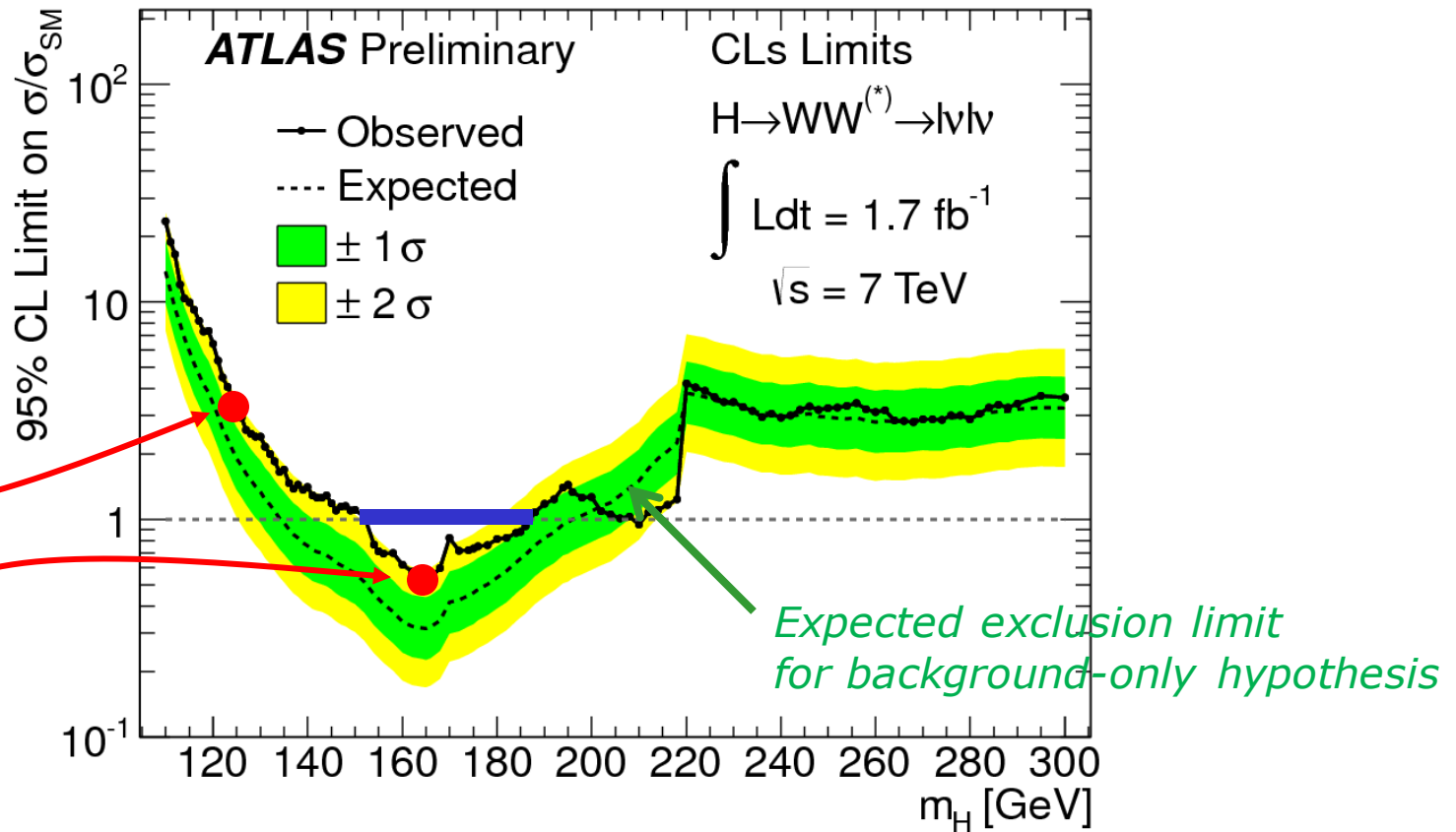


- Construct test statistic
- Obtain expected distributions of q_μ for various μ
 - Determine 'discovery' p-value and signal exclusion limit
- Repeat for each assumed m_H ..



Example – 95% Exclusion limit vs m_H for $H \rightarrow WW$

Example point: $\approx 3 \times$ SM $H \rightarrow WW$ cross-section excluded at $m_H = 125$ GeV

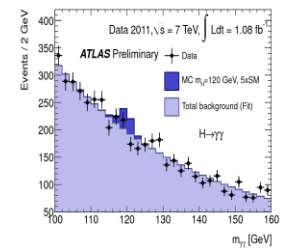
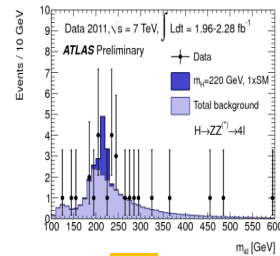
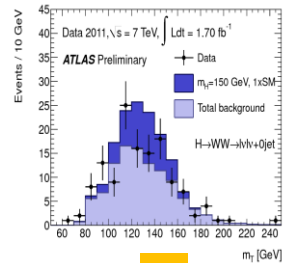


Example point: $\approx 0.5 \times$ SM $H \rightarrow WW$ cross-section excluded at $m_H = 165$ GeV

Higgs with $1.0 \times$ SM cross-section excluded at 95% CL for m_H in range [150, ~187]

Combining Higgs channels (and experiments)

- Procedure: define **joint likelihood**



$$L(\mu, \theta_{comb}) = L_{H \rightarrow WW}(\mu, \theta_{WW}) \cdot L_{H \rightarrow ZZ}(\mu, \theta_{ZZ}) \cdot L_{H \rightarrow \gamma\gamma}(\mu, \theta_{\gamma\gamma}) \cdot \dots$$

$$L(\mu, \theta_{LHC}) = L_{ATLAS}(\mu, \theta_{ATLAS}) \cdot L_{CMS}(\mu, \theta_{CMS}) \cdot \dots$$

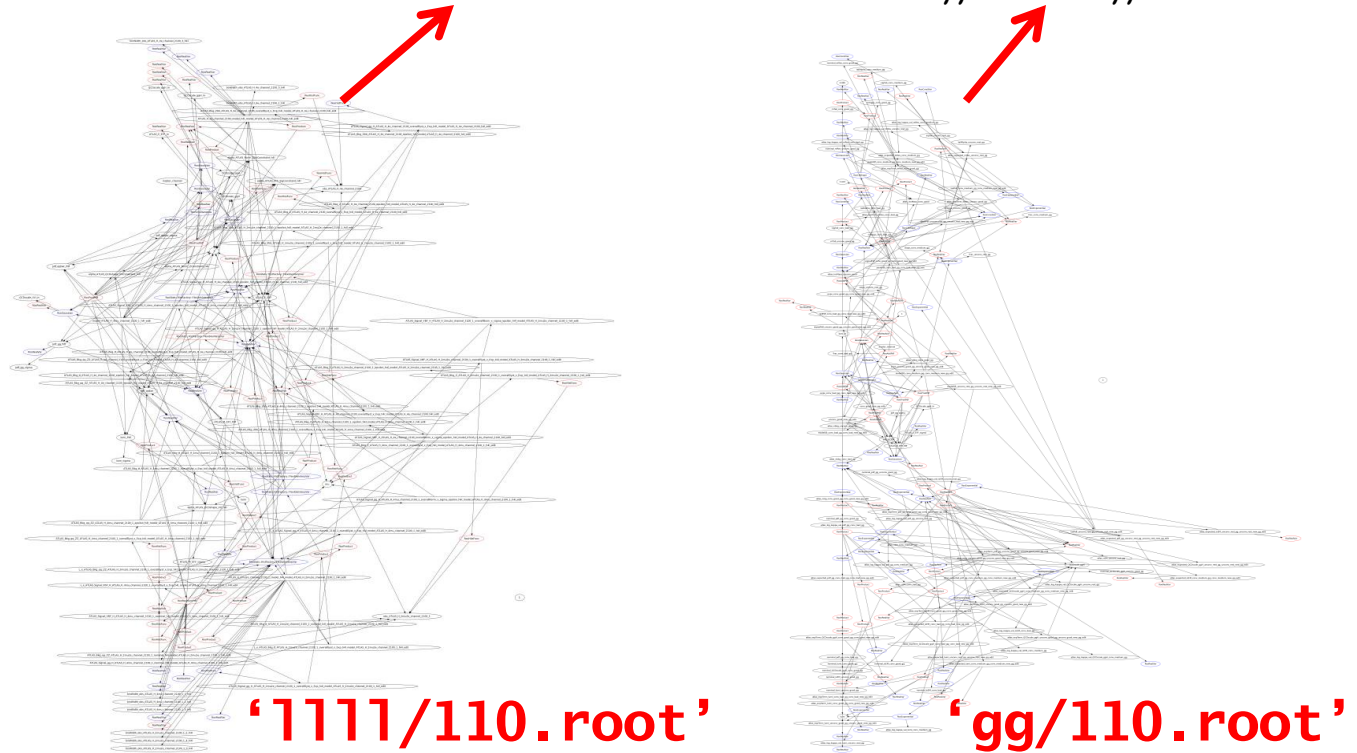
- Correlations between $\theta_{WW}, \theta_{\gamma\gamma}$ etc and between $\theta_{ATLAS}, \theta_{CMS}$ requires careful consideration!**
- The construction profile likelihood ratio test statistic from joint likelihood and proceed as usual

$$\tilde{q}_\mu = -2 \ln \frac{L(data | \mu, \hat{\hat{\theta}}_\mu)}{L(data | \hat{\mu}, \hat{\theta})}$$

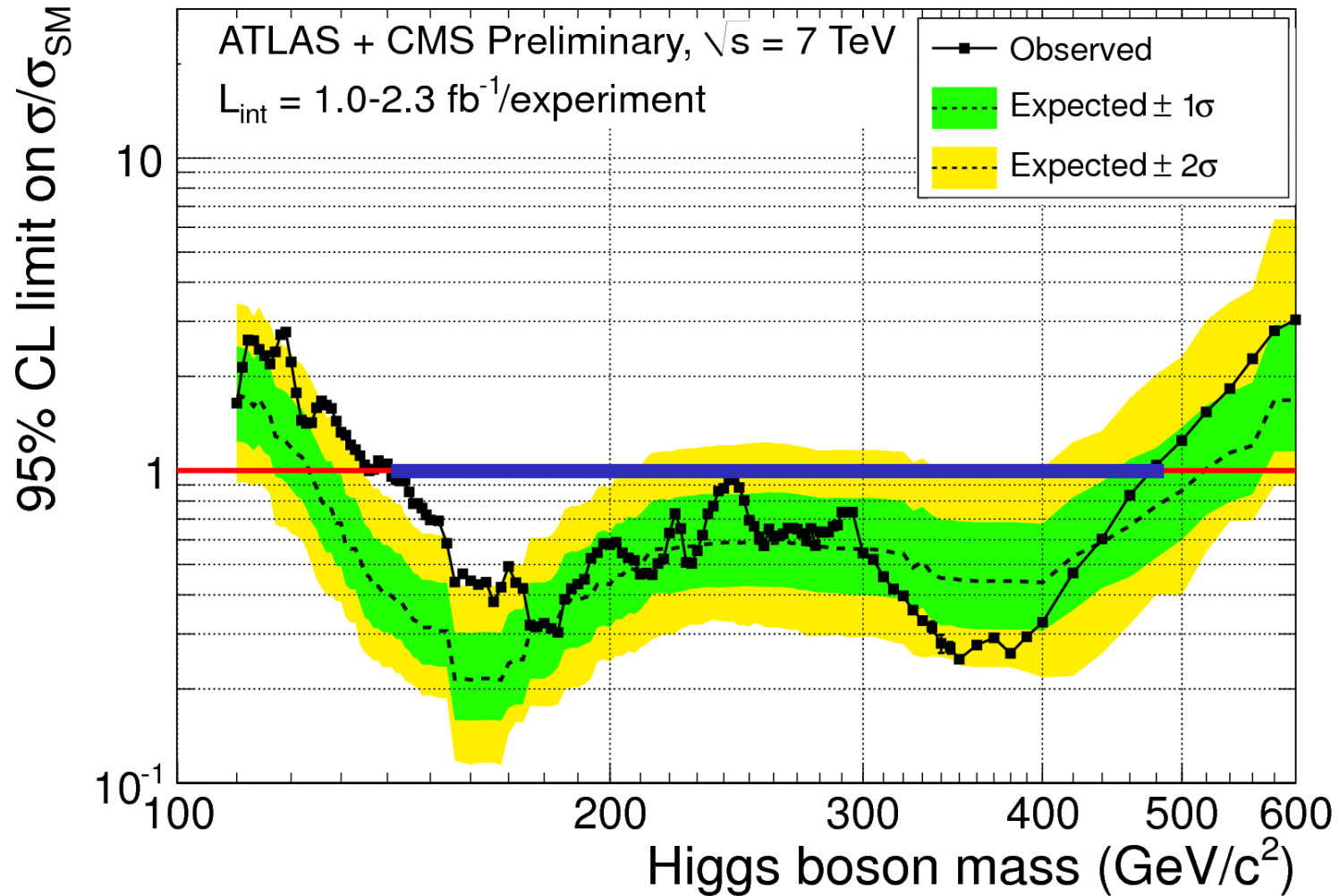
A word on the machinery

- Common tools (RooFit/RooStats - all available in ROOT) have been developed in past 2/3 years to facilitate these combinations
 - Analytical description of likelihood of each component stored and delivered in a uniform language ('RooFit workspaces')
 - Construction of joint likelihood *technically* straightforward

$$L(\mu, \theta_{comb}) = L_{H \rightarrow WW}(\mu, \theta_{WW}) \cdot L_{H \rightarrow ZZ}(\mu, \theta_{ZZ}) \cdot L_{H \rightarrow \gamma\gamma}(\mu, \theta_{\gamma\gamma}) \cdot \dots$$



Example – joint ATLAS/CMS Higgs exclusion limit



Switching from 'exclusion' to 'discovery' formulation

'exclusion'

'discovery'

'likelihood assuming μ signal strength' 'likelihood assuming background only'

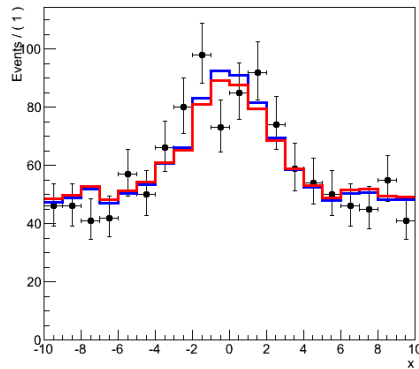
$$\tilde{q}_\mu = -2 \ln \frac{L(\text{data} | \mu, \hat{\theta}_\mu)}{L(\text{data} | \hat{\mu}, \hat{\theta})}$$

$$\tilde{q}_0 = -2 \ln \frac{L(\text{data} | \mu = 0, \hat{\theta}_0)}{L(\text{data} | \hat{\mu}, \hat{\theta})}$$

'likelihood of best fit'

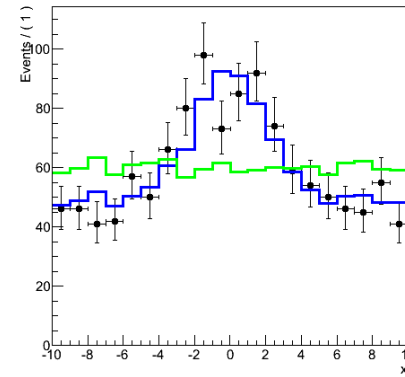
'likelihood of best fit'

$$\tilde{q}_1 = 0.77$$

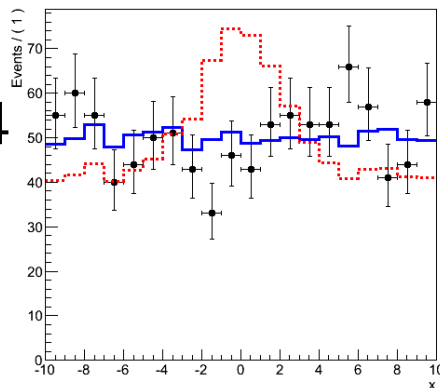


*simulated
data with
signal+bkg*

$$\tilde{q}_0 = 34.7$$

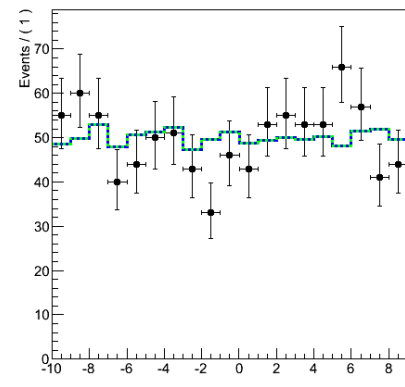


$$\tilde{q}_1 = 52.34$$



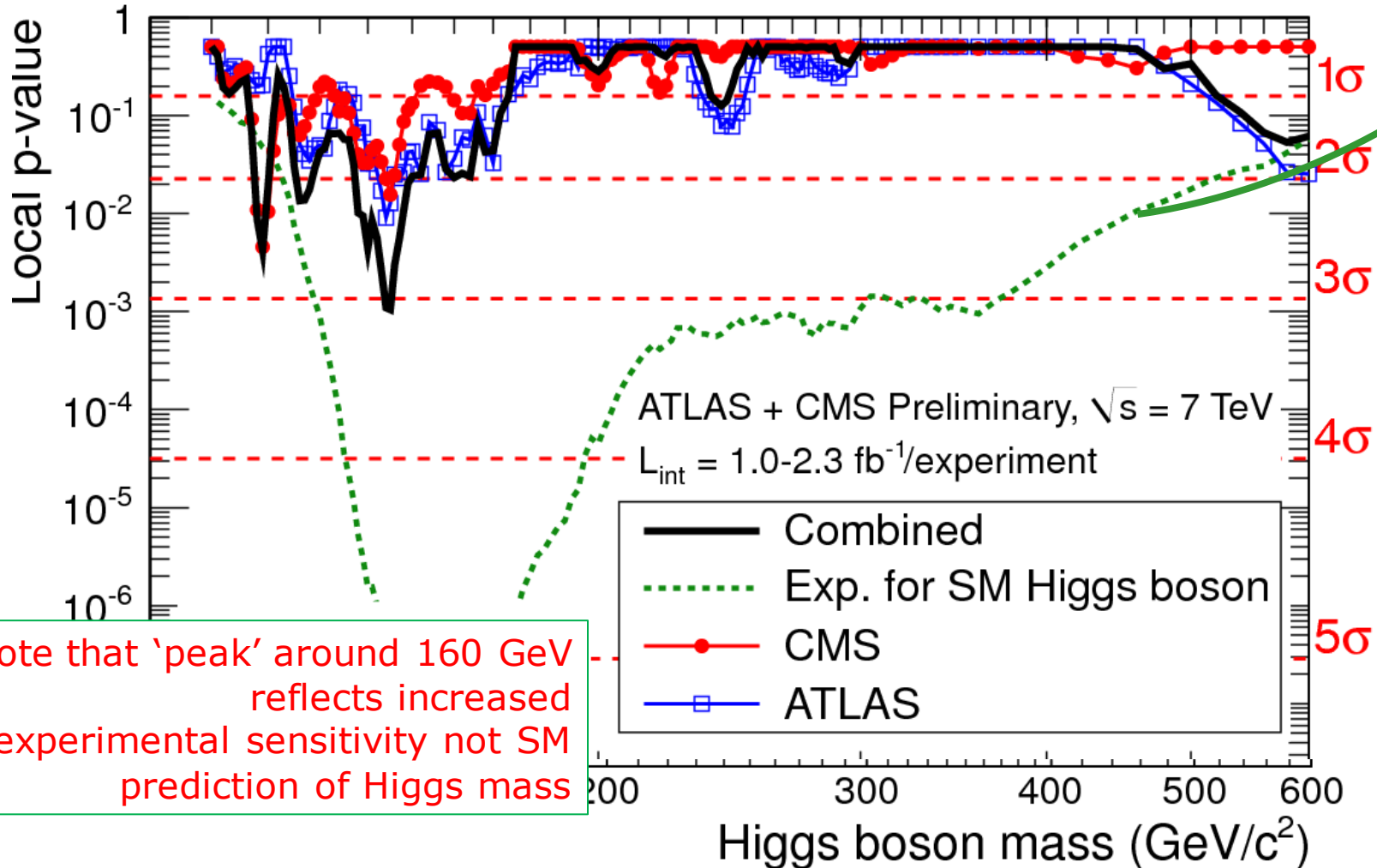
*simulated
data with
bkg only*

$$\tilde{q}_0 = 0$$



Comb: p-value of background-only hypothesis ('discovery')

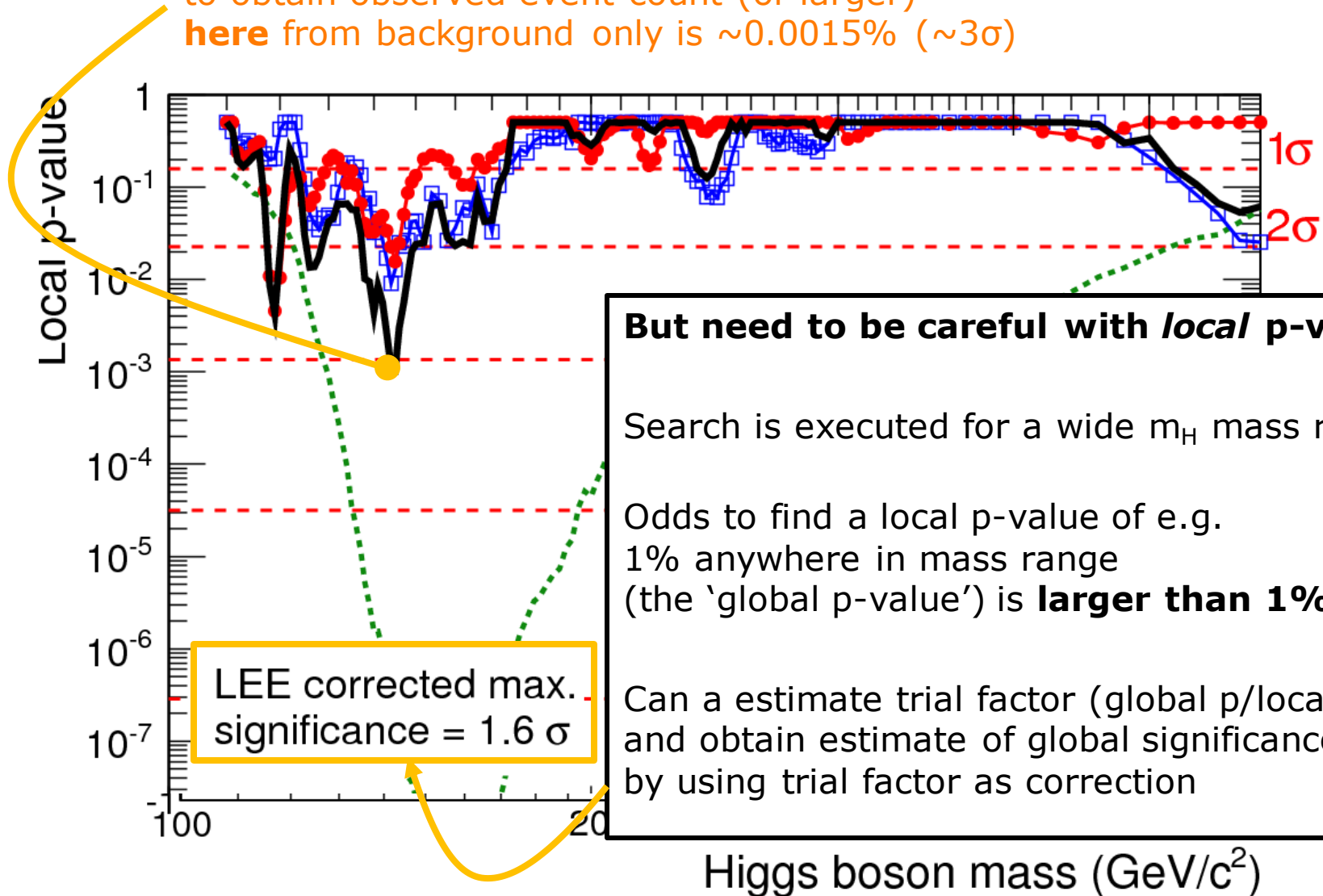
Expected p-value for background hypothesis of a data sample containing SM Higgs boson



Note that 'peak' around 160 GeV reflects increased experimental sensitivity not SM prediction of Higgs mass

Comb: p-value of background-only hypothesis ('discovery')

Example point: at $m_H \approx 150$ GeV probability to obtain observed event count (or larger) **here** from background only is $\sim 0.0015\%$ ($\sim 3\sigma$)



Conclusions

- We are early awaiting more data!