

Search and Discovery Statistics in HEP Lecture 3

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This presentation would have not been possible without the tremendous
help of
the following people throughout many years

Louis Lyons, Alex Read, Bob Cousins, Glen Cowan, Kyle Cranmer
Ofer Vitells & Jonathan Shlomi



What can you expect from the Lectures



Lecture 1: Basic Concepts

Histograms, PDF, Testing Hypotheses,
LR as a Test Statistics, p-value, POWER, CLs
Measurements



Lecture 2: Wald Theorem, Asymptotic Formalism, Asimov Data
Set, Feldman-Cousins, PL & CLs, Asimov Significance



Lecture 3: Look Elsewhere Effect

1D LEE the non-intuitive thumb rule
(upcrossings, trial #~Z)
2D LEE (Euler Characteristic)



Look Elsewhere Effect



E.G., O. Vitells “Trial factors for the look elsewhere effect in high energy physics”,

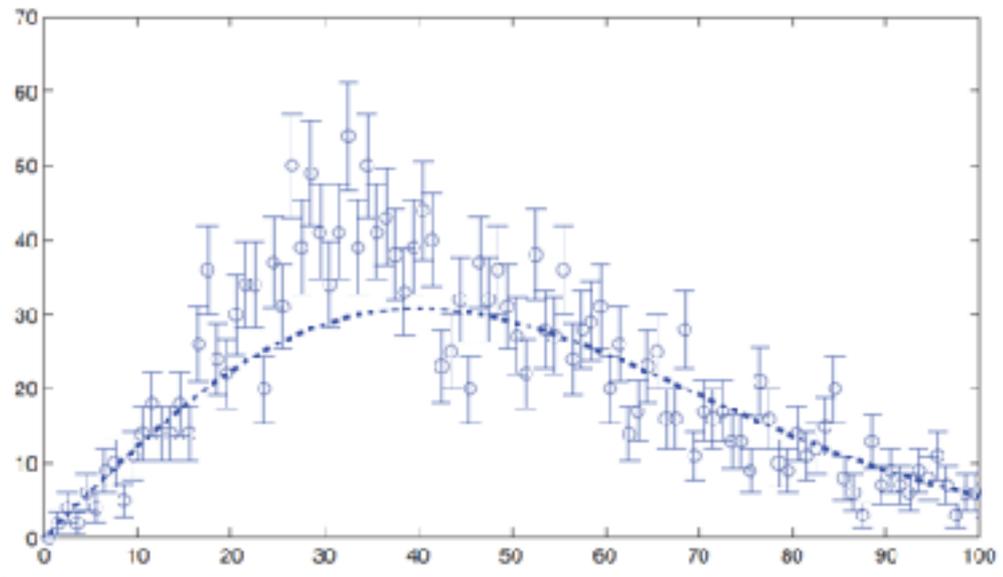
Eur. Phys. J. C 70 (2010) 525

O. Vitells and E. G., Estimating the significance of a signal in a multi-dimensional search,

1669 Astropart. Phys. 35 (2011) 230, arXiv:1105.4355

Look Elsewhere Effect

- Is there a signal here?

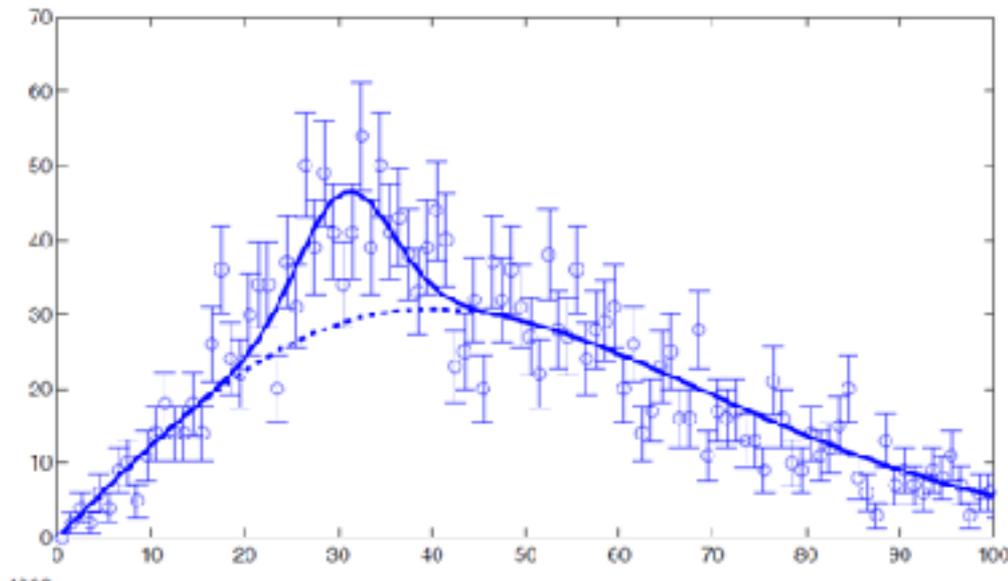


Look Elsewhere Effect

- Looks like a signal at $m=30$
- What is its significance?

Test the BG hypothesis
At $m=30$

$$q_0(\theta) = \begin{cases} -2 \log \frac{L(\mu = 0)}{L(\hat{\mu}, \theta)} \\ 0 \end{cases}$$



$$q_{fix,obs} = -2 \ln \frac{L(b)}{L(\hat{\mu}_s(m=30) + b)}$$

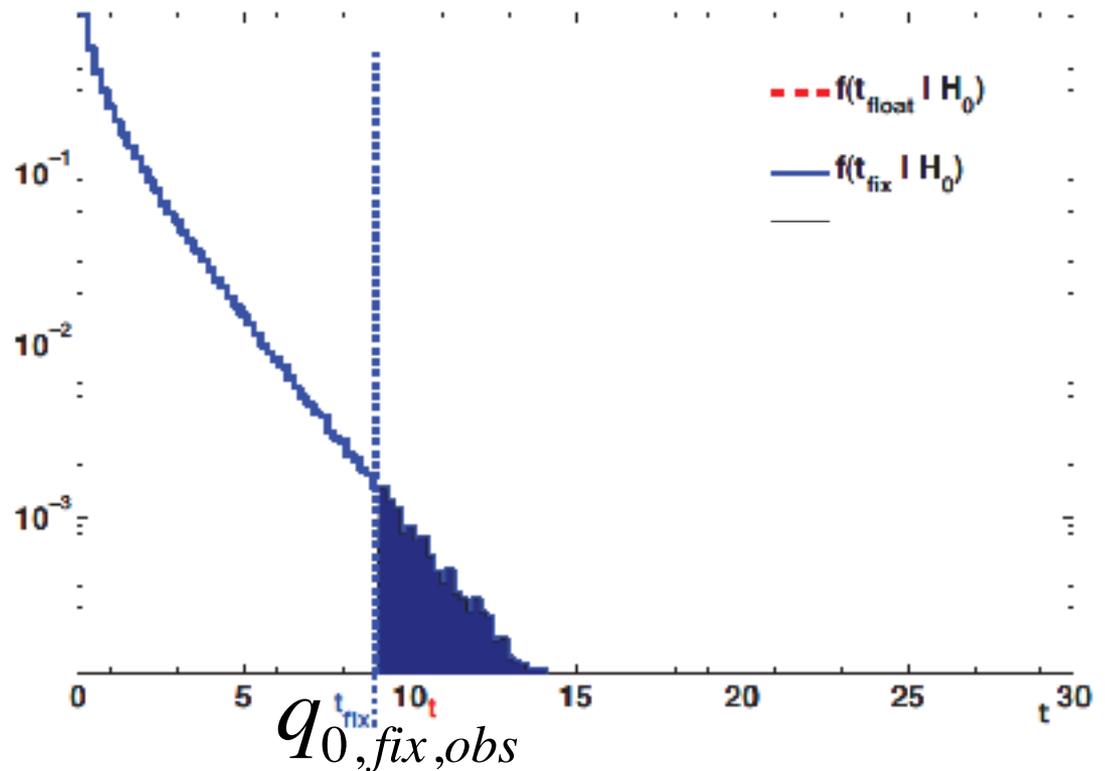
$$Z = \sqrt{q_{0,fix,obs}}$$

Look Elsewhere Effect

$$q_{0,fix} = -2 \ln \frac{L(\mu = 0)}{L(\hat{\mu}_s(30) + b)}$$

$$f(q_{0,fix} | H_0) \sim \chi^2$$

$$p_{fix} = \int_{q_{fix,obs}}^{\infty} f(q_0 | H_0) dq_0$$

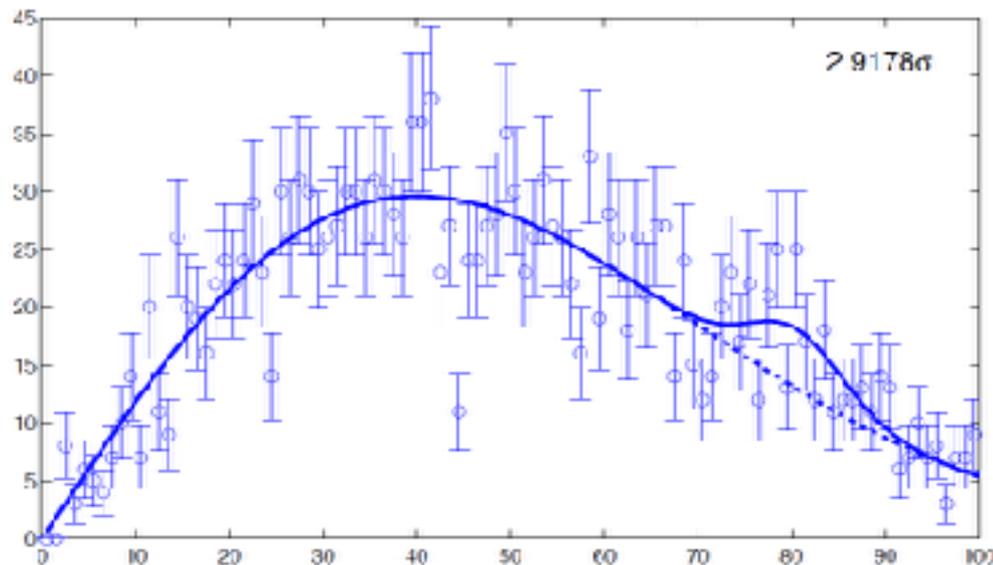


p_{fix} answers the question :

What is the probability to have a fluctuation as or bigger than the observed one?

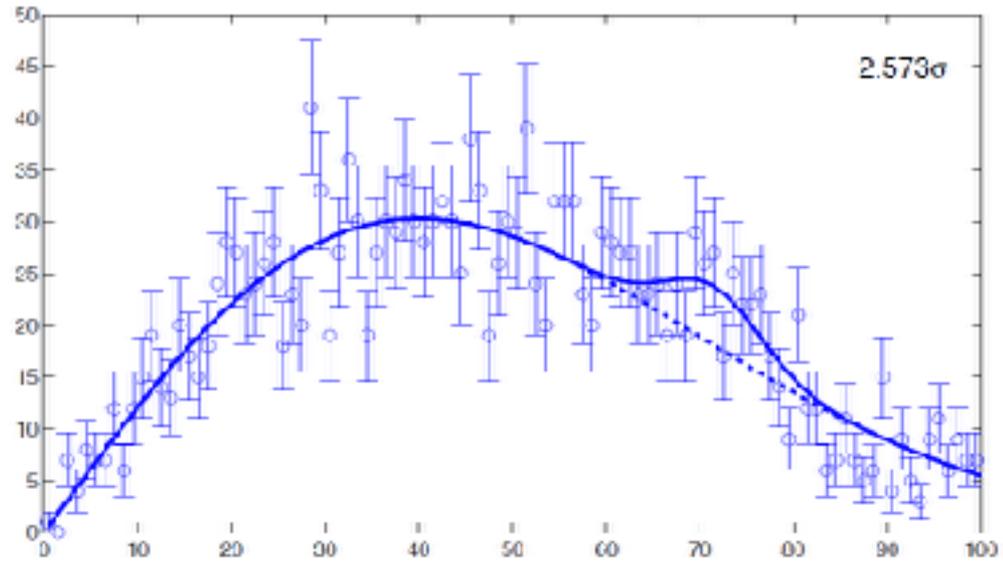
Look Elsewhere Effect

- Would you ignore this signal, had you seen it?



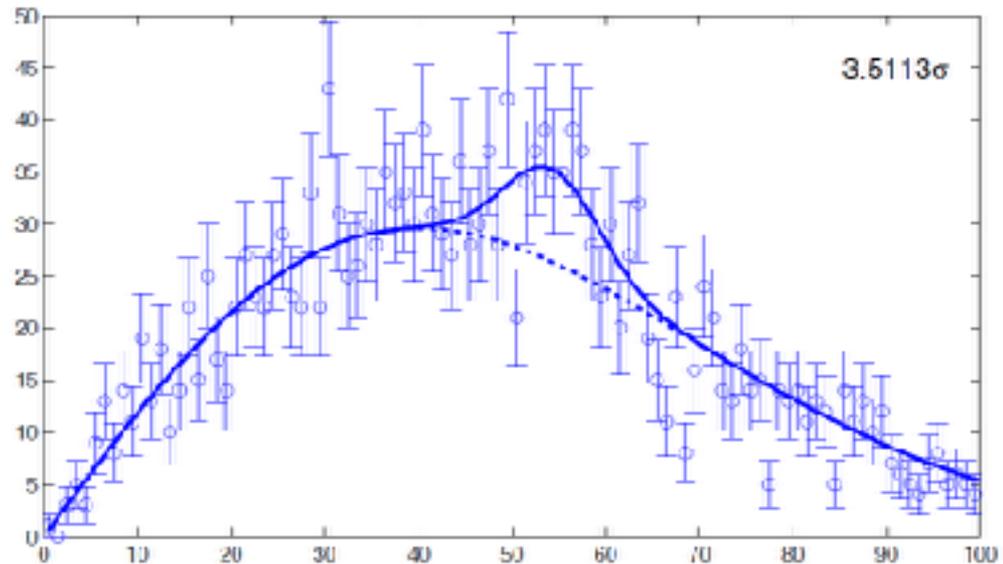
Look Elsewhere Effect

•Or this?



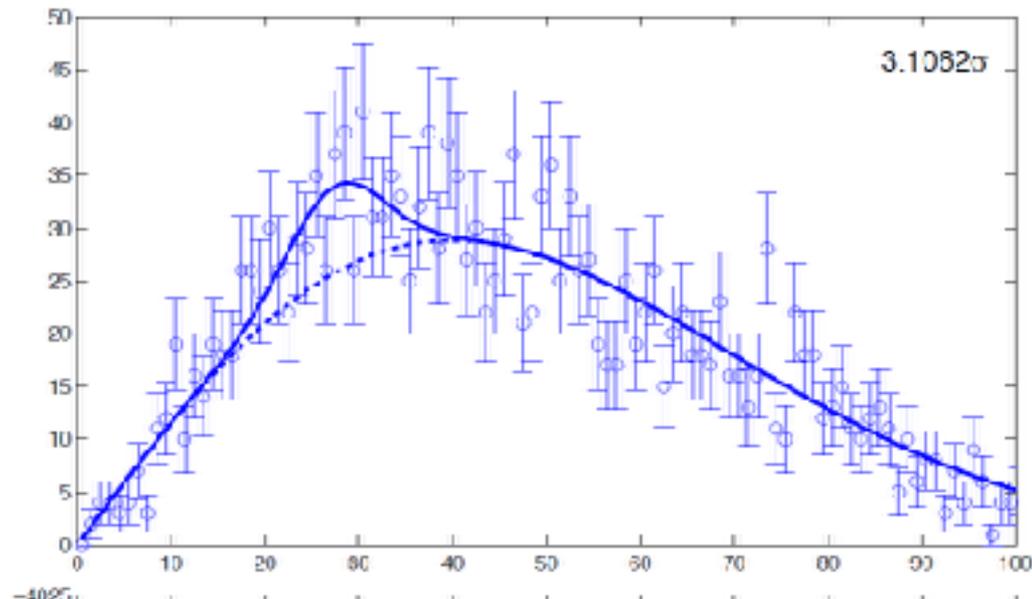
Look Elsewhere Effect

•Or this?



Look Elsewhere Effect

- Or this?
- Obviously NOT!
- ALL THESE "SIGNALS" ARE BG FLUCTUATIONS



The right question :

What is the probability to have a fluctuation as or bigger than the observed one

***ANYWHERE** in the mass search range?*

Look Elsewhere Effect

- Having no idea where the signal might be there are two equivalent options

- **OPTION I:**

scan the mass range in pre-defined steps and test any disturbing fluctuations

Perform a sliding window fixed mass analysis

$$q_{0, \text{float}} = \max_m (q_0(m))$$

$$P_{\text{float}} = \int_{q_{\text{float, obs}}}^{\infty} f(q_{0, \text{float}} | H_0) dq_{0, \text{float}}$$



- **OPTION II:**

Perform a floating mass analysis

$$q_{0, \text{float}} = q_0(\hat{m}) = -2 \ln \frac{L(b)}{L(\hat{\mu}_s(\hat{m}) + b)}$$

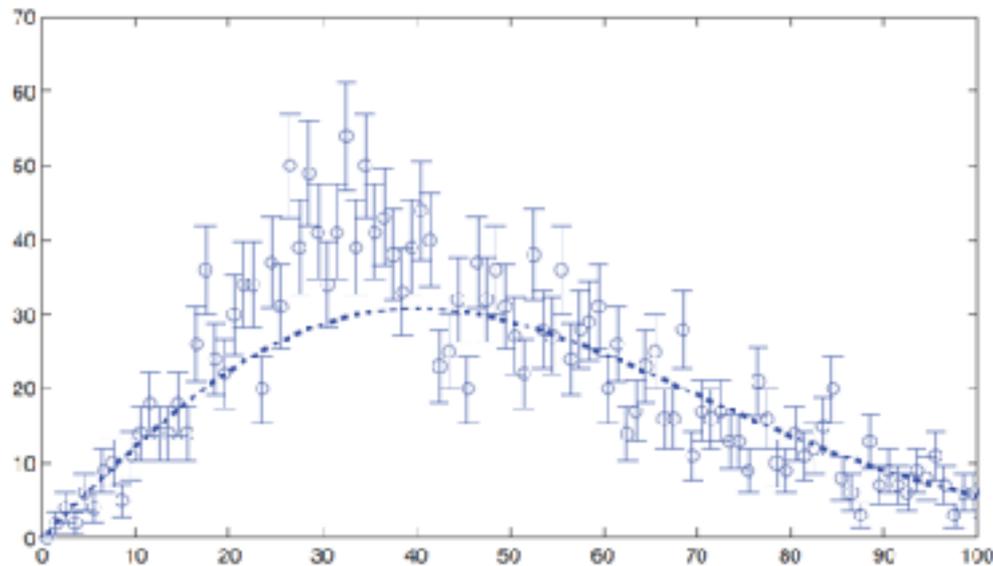
$$P_{\text{float}} = \int_{q_{\text{float, obs}}}^{\infty} f(q_{0, \text{float}} | H_0) dq_{0, \text{float}}$$



Sliding Window

- Scan and perform a fixed mass analysis at each point

$$q_0 = -2 \ln \frac{L(\mu = 0)}{L(\hat{\mu}_s(m) + b)}$$



- The scan resolution must be less than the signal mass resolution

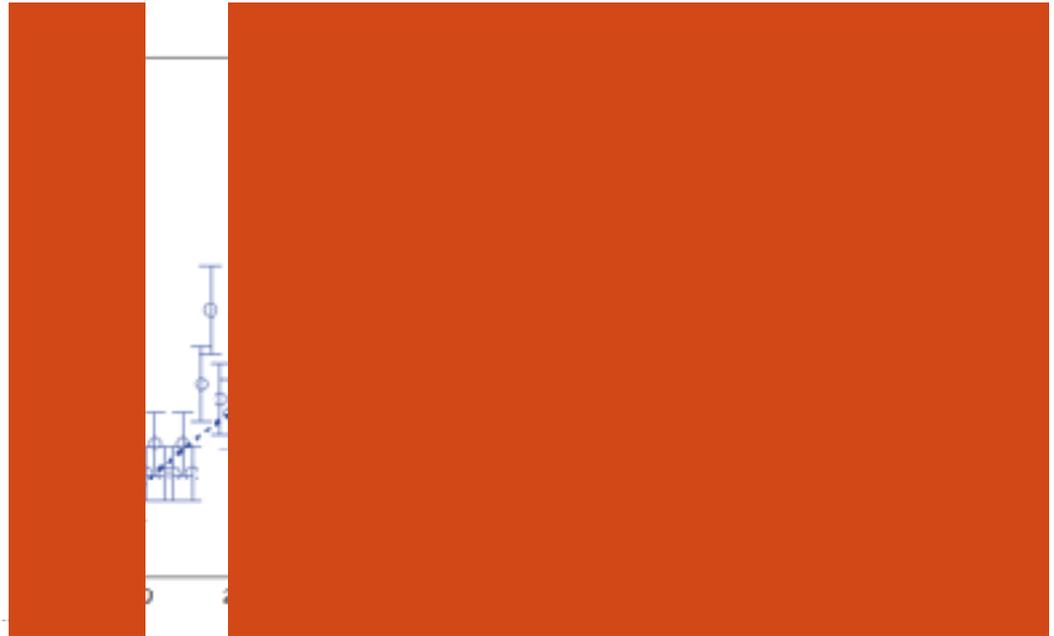
Sliding Window

$$q_0 = -2 \ln \frac{L(\mu = 0)}{L(\hat{\mu}_s(m) + b)}$$



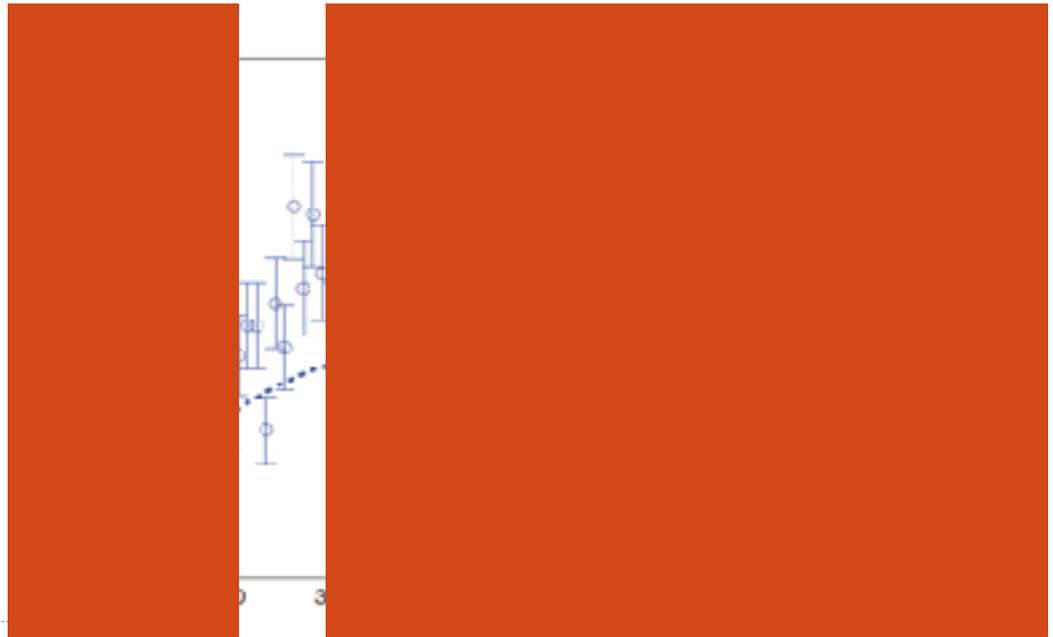
Sliding Window

$$q_0 = -2\ln \frac{L(\mu = 0)}{L(\hat{\mu}_s(m) + b)}$$



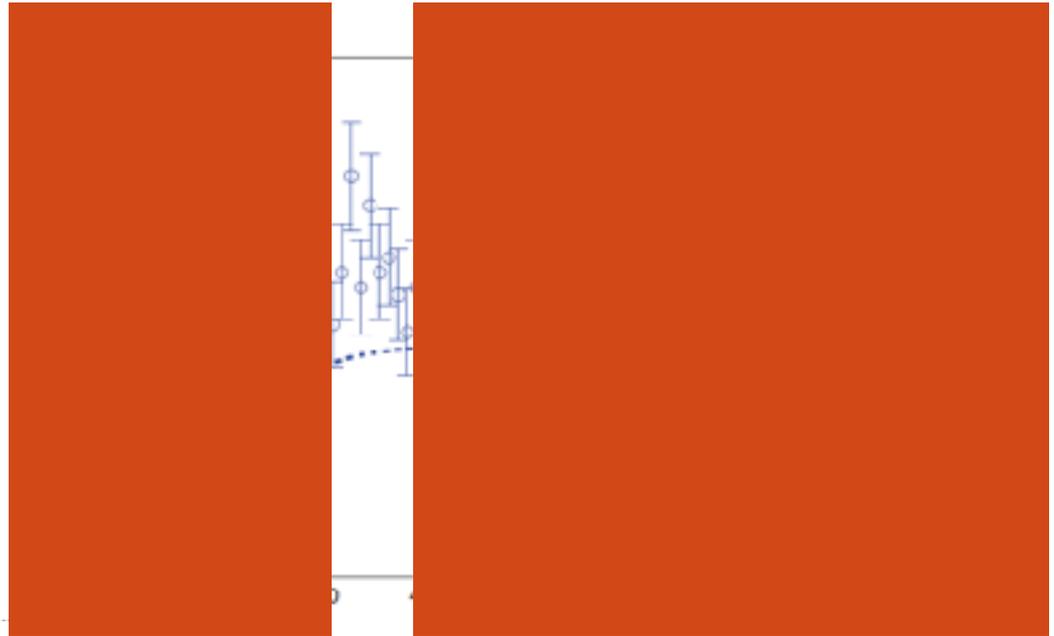
Sliding Window

$$q_0 = -2\ln \frac{L(\mu = 0)}{L(\hat{\mu}_s(m) + b)}$$



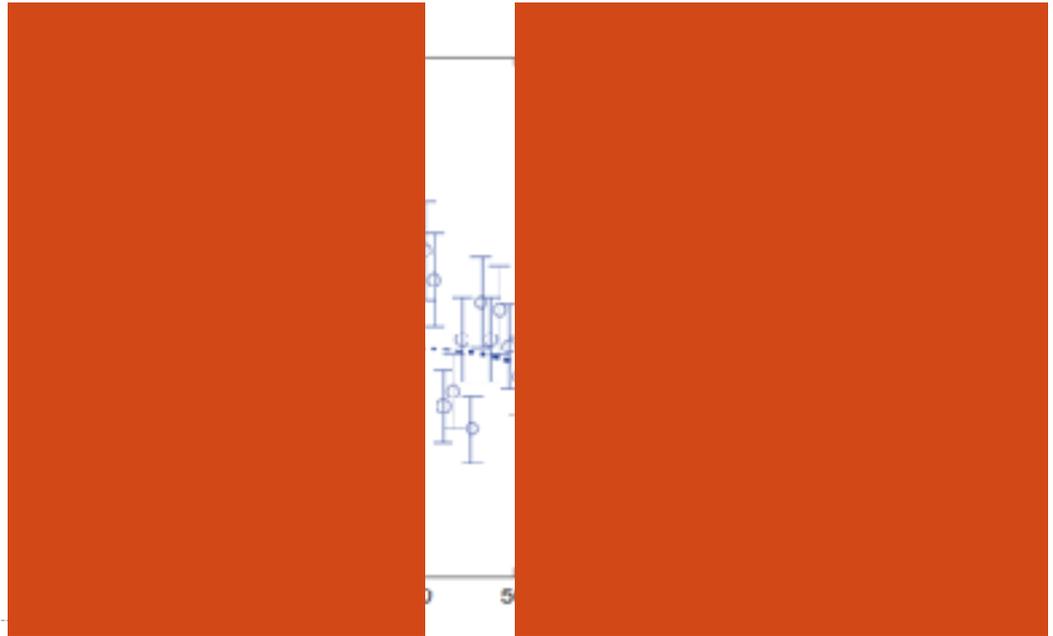
Sliding Window

$$q_0 = -2\ln \frac{L(\mu = 0)}{L(\hat{\mu}_s(m) + b)}$$



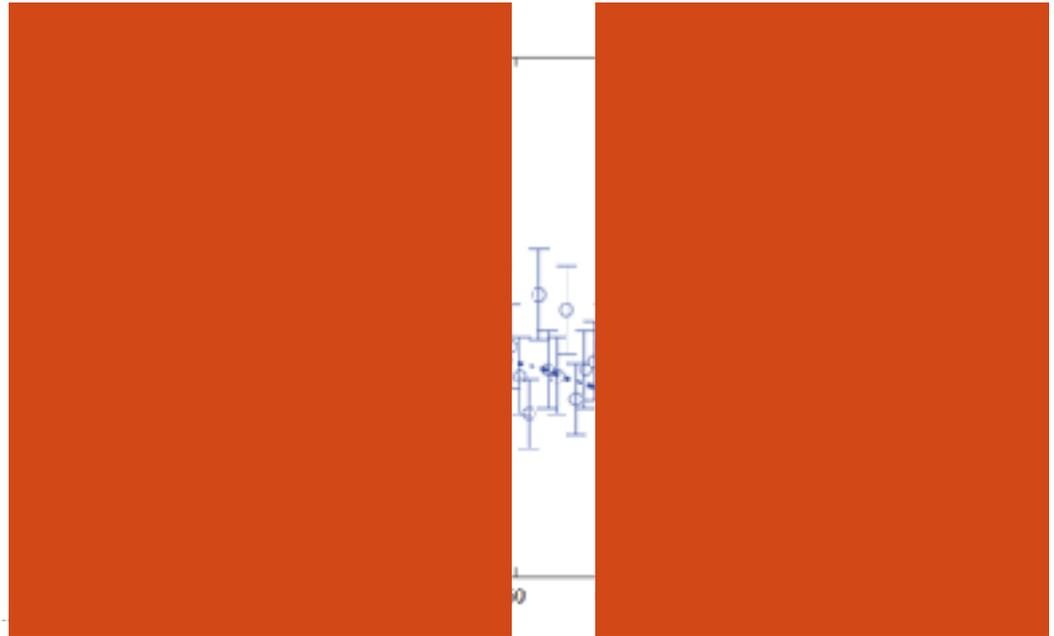
Sliding Window

$$q_0 = -2 \ln \frac{L(\mu = 0)}{L(\hat{\mu}_s(m) + b)}$$



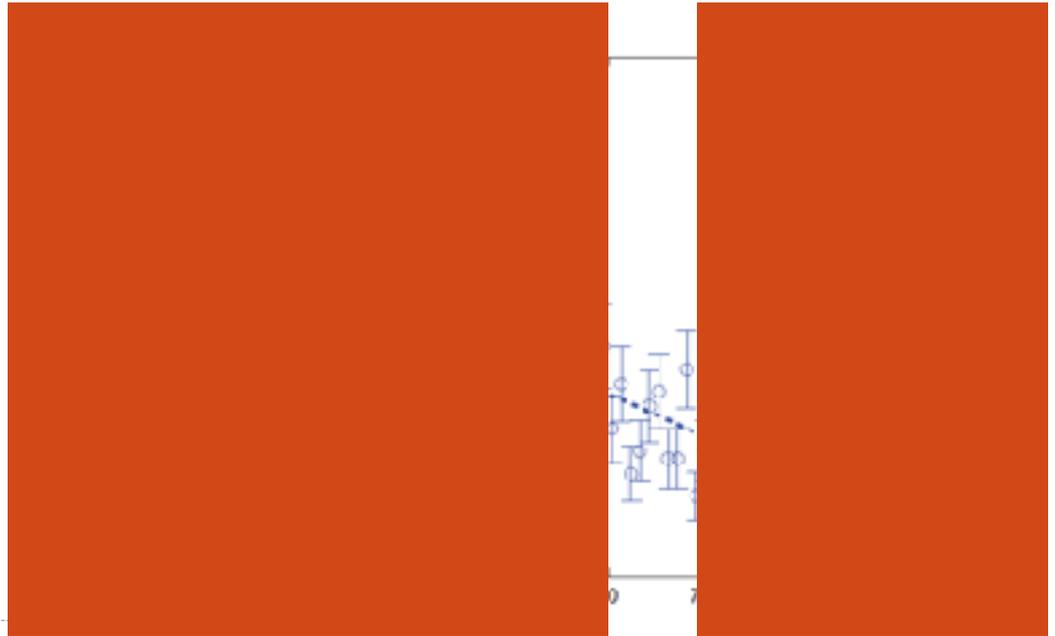
Sliding Window

$$q_0 = -2\ln \frac{L(\mu = 0)}{L(\hat{\mu}_s(m) + b)}$$



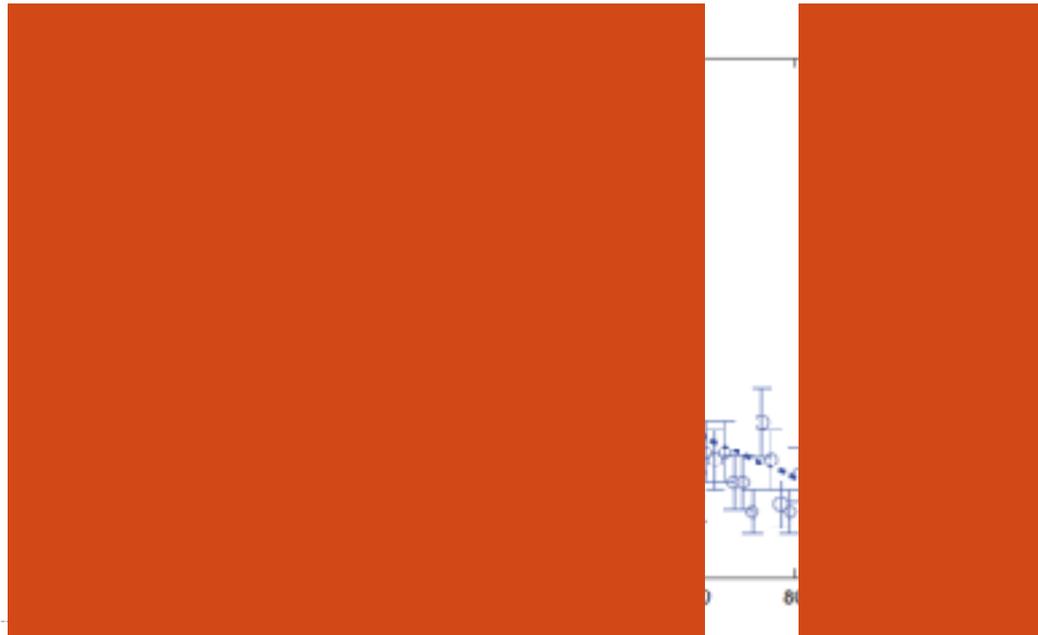
Sliding Window

$$q_0 = -2 \ln \frac{L(\mu = 0)}{L(\hat{\mu}_s(m) + b)}$$



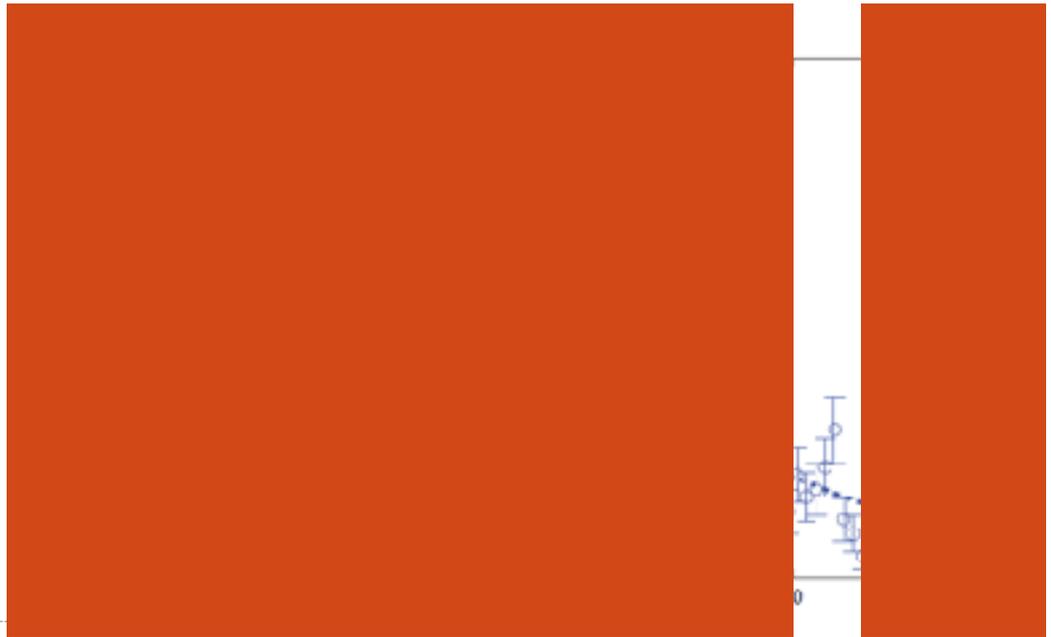
Sliding Window

$$q_0 = -2\ln \frac{L(\mu = 0)}{L(\hat{\mu}_s(m) + b)}$$



Sliding Window

$$q_0 = -2\ln \frac{L(\mu = 0)}{L(\hat{\mu}_s(m) + b)}$$



Sliding Window

- Assuming the signal can be only at one place
- pick the one with the **MAXIMUM SIGNIFICANCE**



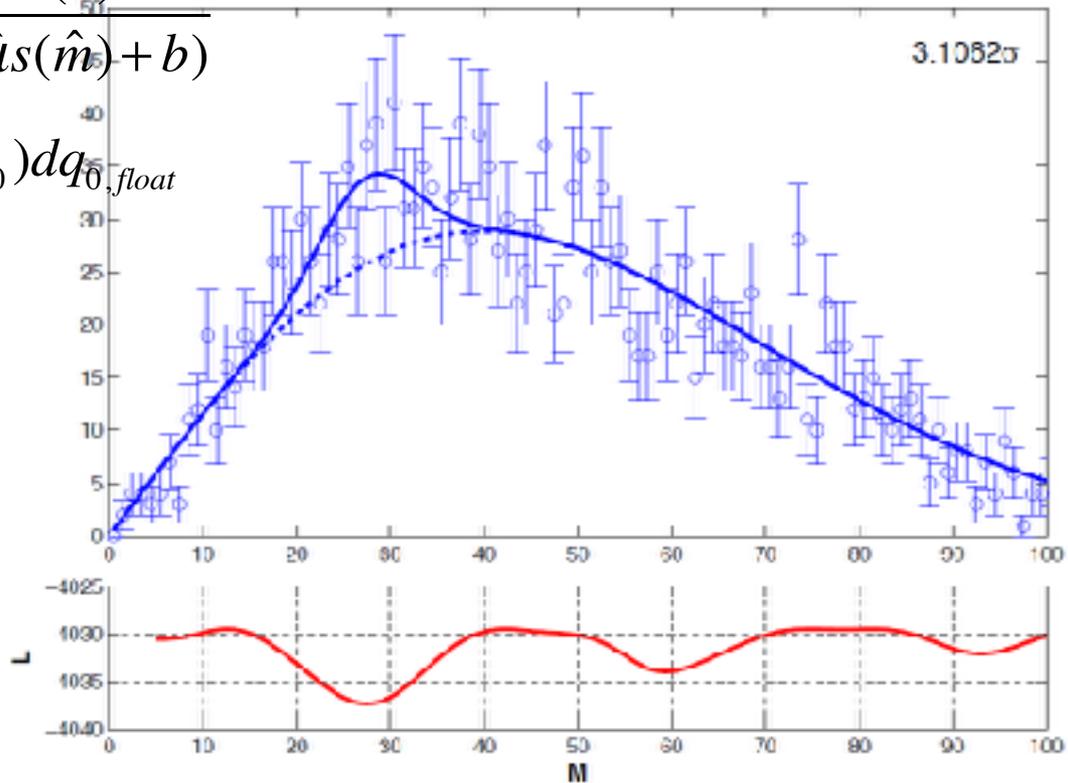
$$q_{0, float} = \max_m (q_0(m))$$

Look Elsewhere Effect: Floating Mass

OPTION II

$$q_{0, \text{float}} = q_0(\hat{m}) = -2 \ln \frac{L(b)}{L(\hat{\mu}_s(\hat{m}) + b)}$$

$$P_{\text{float}} = \int_{q_{\text{float, obs}}}^{\infty} f(q_{0, \text{float}} | H_0) dq_{0, \text{float}}$$

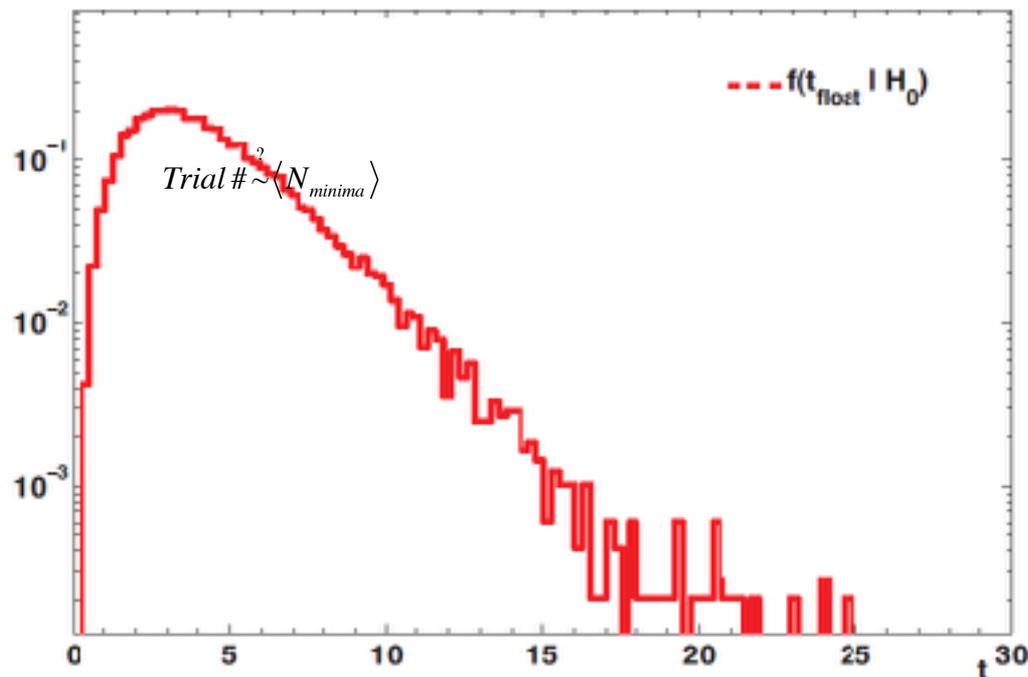


Look Elsewhere Effect

- The distribution $f(q_{\text{float}}|H_0)$ does not follow a chi-squared with 2 dof because the mass parameter is not defined under the null hypothesis

$$\exists m_{\text{fix}} \quad q_0(\hat{m}) \geq q_0(m_{\text{fix}})$$

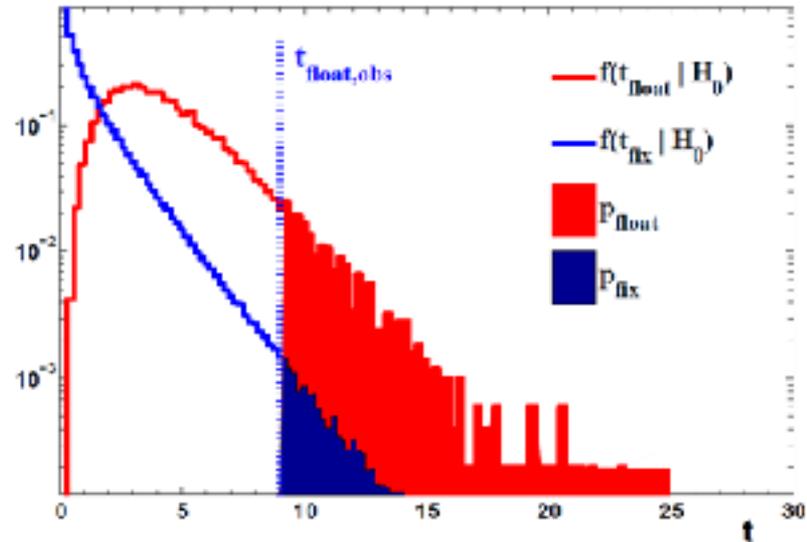
The χ^2_1 distribution is pushed to the right



trial#

- Assume a maximal local fluctuation at mass $\hat{m} = 30$
- The observed q_0 is given by

$$q_{0,obs} = -2 \ln \frac{L(\mu = 0)}{L(\hat{\mu}_s(m) + b)}$$



$$P_{fix} = \int_{q_{0,obs}}^{\infty} f(q_{0,fix} | H_0) dq_{0,fix}$$

$$P_{float} = \int_{q_{0,obs}}^{\infty} f(q_{0,float} | H_0) dq_{0,float}$$

$$trial \# = \frac{P_{float}}{P_{fix}}$$

Can we calculate analytically the floating mass p-value

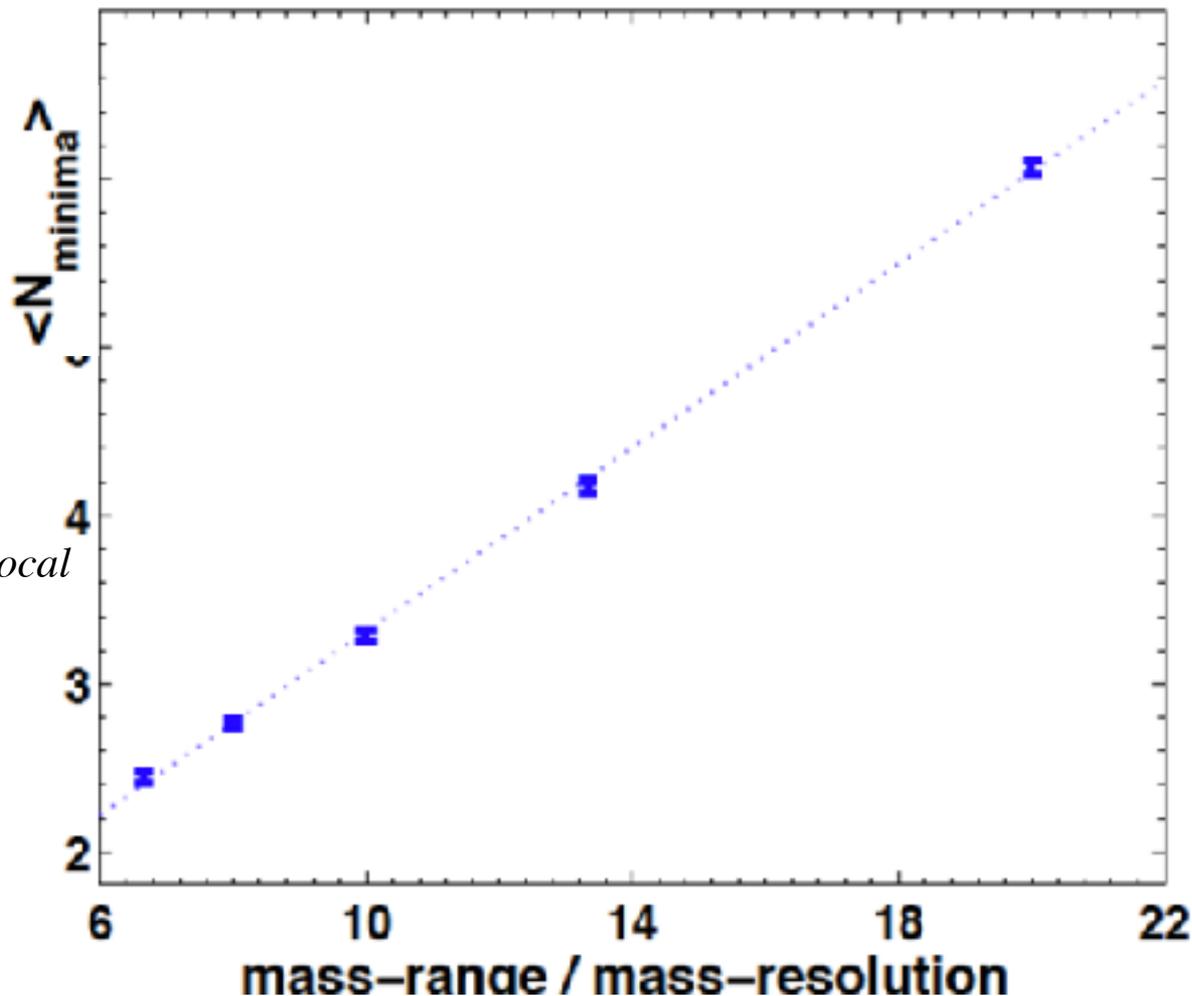
(Wrong) Thumb Rule

$$\langle N_{\text{minima}} \rangle \sim \frac{\text{Mass Range}}{\text{Mass Resolution}}$$

$$\text{Trial \#} \sim \langle N_{\text{minima}} \rangle$$

$$\text{Trial \#} \stackrel{?}{=} \langle N_{\text{minima}} \rangle P_{\text{local}}$$

The answer is NO



The right question :

*What is the probability to have a fluctuation
as or bigger than the observed one*

***ANYWHERE** in the mass search range?*

*Let θ be a nuisance parameter
undefined under the null hypothesis.*

Define $q(\hat{\theta}) = \max_{\theta} (q(\theta))$

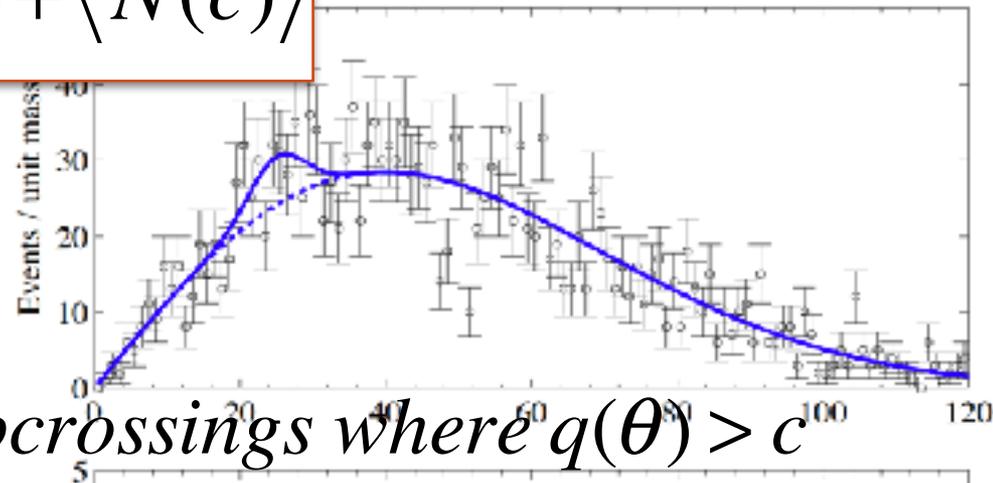
Davies (1987) finds, for $c \gg 1$

$P(q(\hat{\theta}) > c) \sim P(\chi_1^2 > c) + \langle N(c) \rangle$

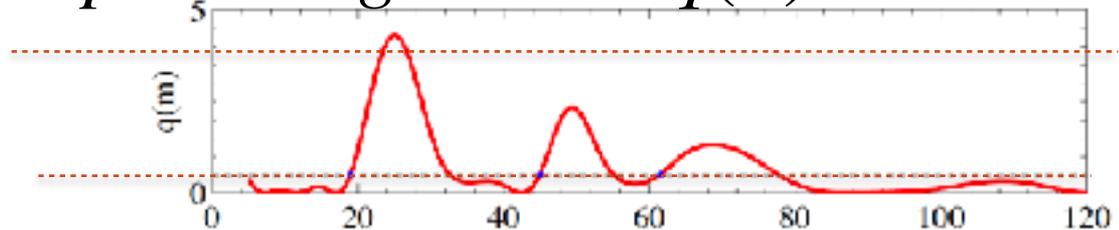
*$\langle N(c) \rangle =$ Number of
upcrossings $q(\theta) > c$*

Davies Formula

$$P(q(\hat{\theta}) > c) \sim P(\chi_1^2 > c) + \langle N(c) \rangle$$

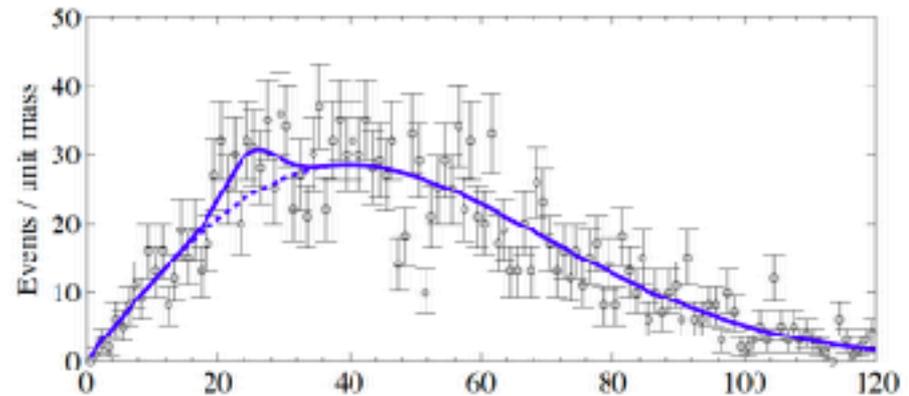


$\langle N(c) \rangle =$ Number of upcrossings where $q(\theta) > c$



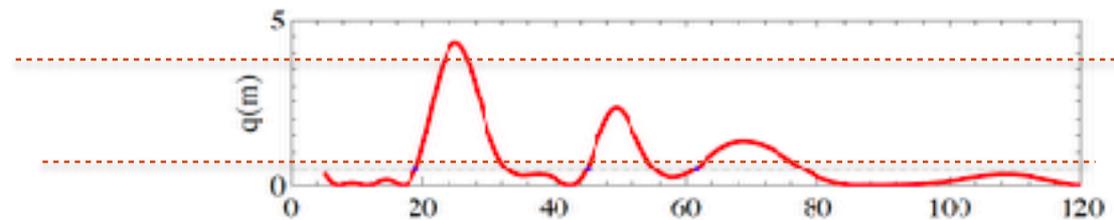
for $c \gg 1 \rightarrow \langle N(c) \rangle \ll 1$

Making Davies Formula Accessible



$$\langle N(c) \rangle \ll 1$$

$$\langle N(c) \rangle \sim e^{-c/2}$$



$$P(q(\hat{\theta}) > c) \sim P(\chi_1^2 > c) + \langle N(c_0) \rangle \frac{\langle N(c) \rangle}{\langle N(c_0) \rangle}$$

$$P(q(\hat{\theta}) > c) \sim P(\chi_1^2 > c) + \langle N(c_0) \rangle e^{-(c-c_0)/2}$$

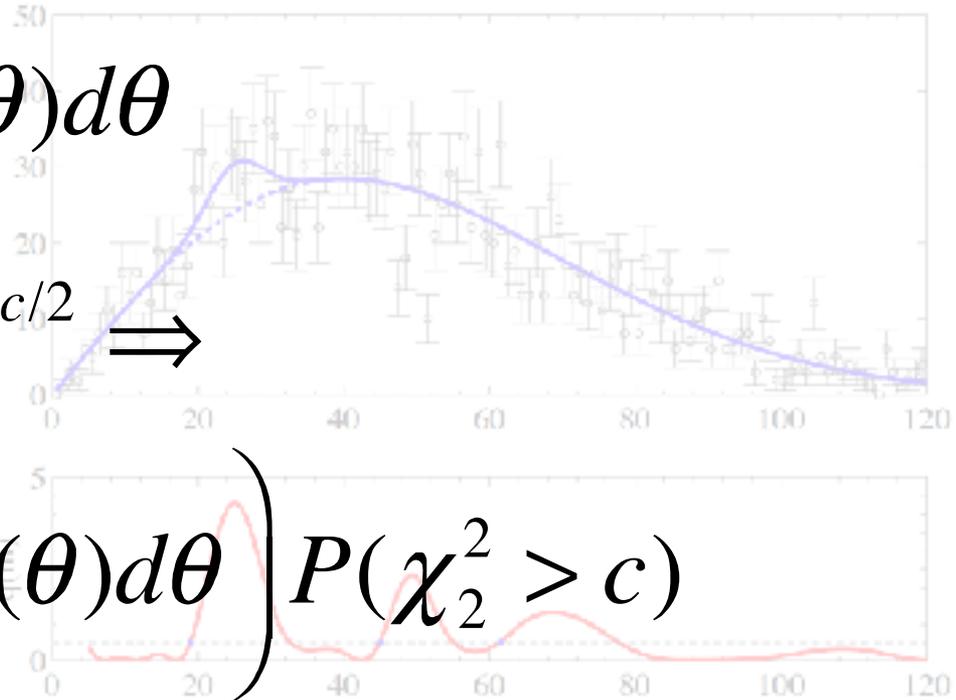
Davies Formula

$$\langle N(c) \rangle \approx \frac{e^{-c/2}}{\sqrt{2\pi}} \int_{\theta} C(\theta) d\theta$$

$$P(\chi_2^2 > c) \xrightarrow{c \rightarrow \infty} e^{-c/2} \Rightarrow$$

$$\langle N(c) \rangle \approx \left(\frac{1}{\sqrt{2\pi}} \int_{\theta} C(\theta) d\theta \right) P(\chi_2^2 > c)$$

$$\langle N(c) \rangle = \mathcal{N} P(\chi_2^2 > c)$$



$$P(q(\hat{\theta}) > c) \sim P(\chi_1^2 > c) + \mathcal{N} P(\chi_2^2 > c)$$



Trial

$$P(\chi_1^2 > c) \xrightarrow{c \gg 1} \sqrt{\frac{2}{c}} \frac{e^{-c/2}}{\Gamma\left(\frac{1}{2}\right)}$$
$$P(\chi_2^2 > c) \xrightarrow{c \gg 1} e^{-c/2}$$

$$\text{trial \#} = \frac{P(q(\hat{\theta}) > c)}{P(q(\theta) > c)} \approx$$

$$\approx 1 + \mathcal{N} \frac{P(\chi_2^2 > c)}{P(\chi_1^2 > c)} \Rightarrow$$

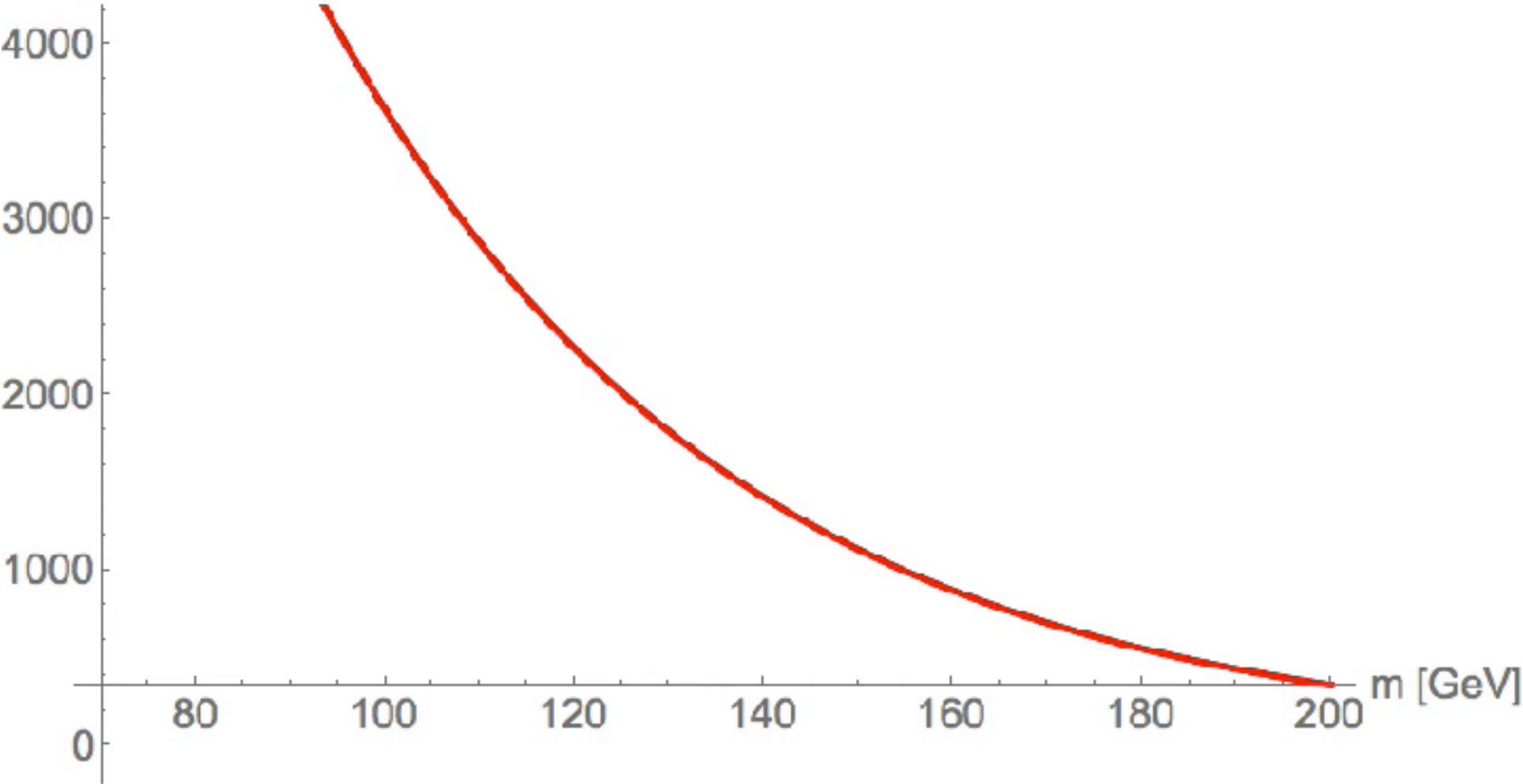
$$\text{trial \#} \approx 1 + \mathcal{N} \sqrt{\frac{c}{2}} \Gamma(1/2) \Rightarrow$$

$$\text{trial \#} \approx 1 + \sqrt{\frac{\pi}{2}} \mathcal{N} Z_{fix}$$

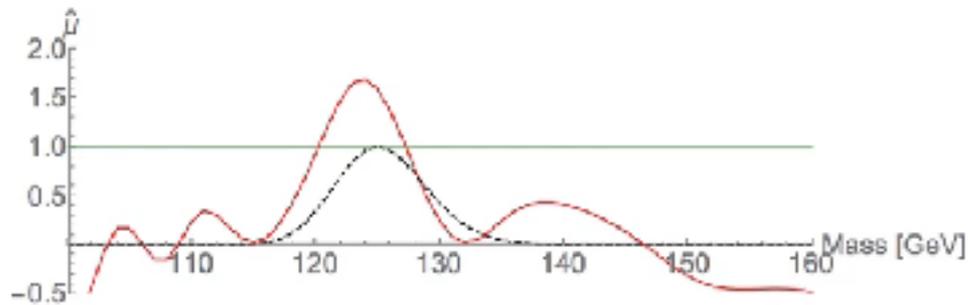
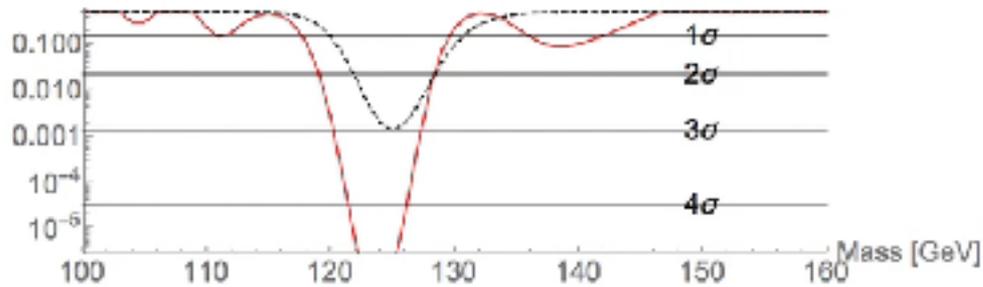
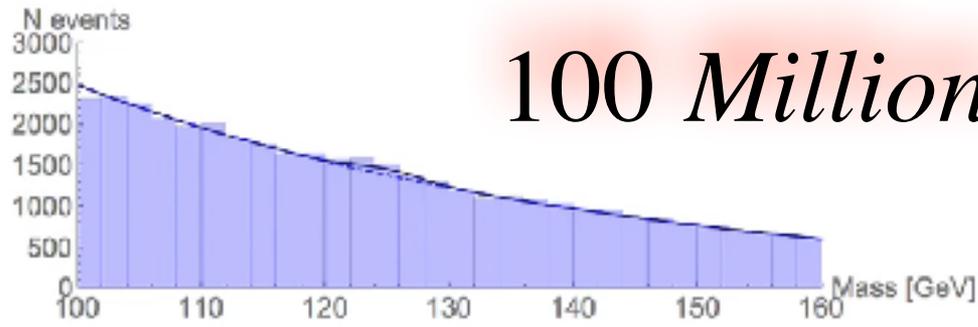
What is Going On?



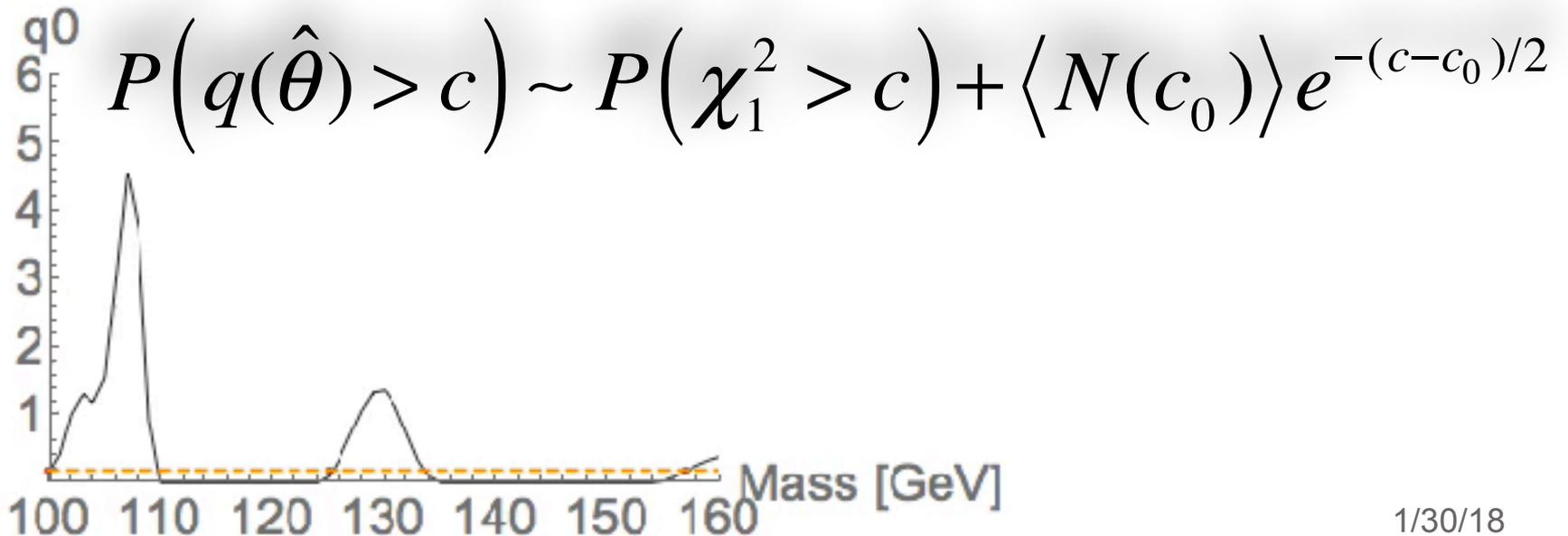
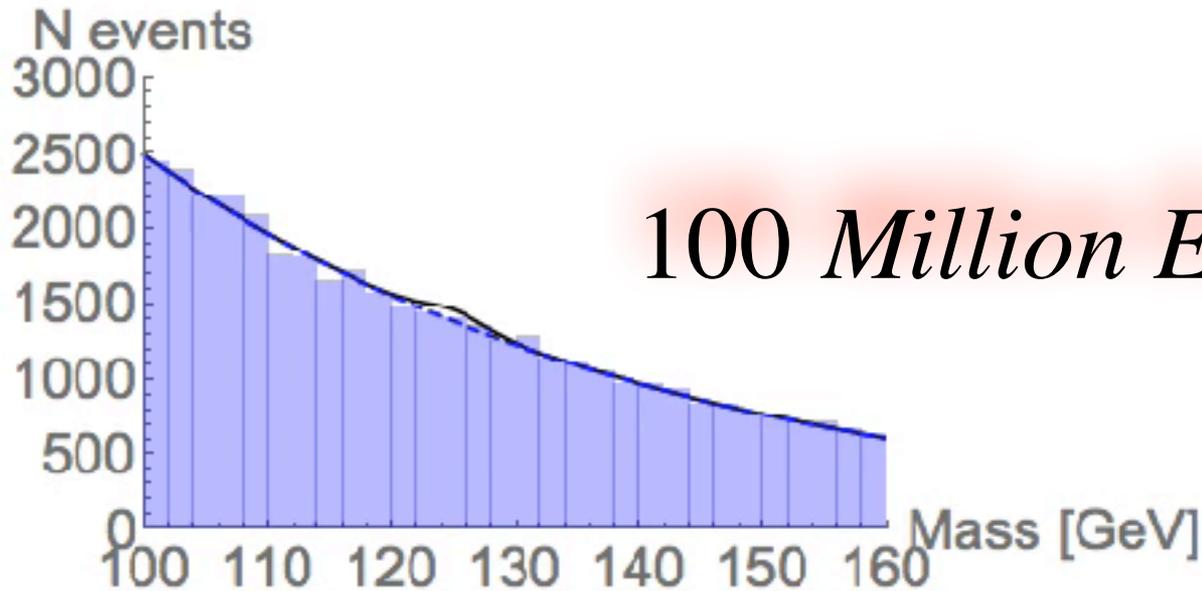
N events

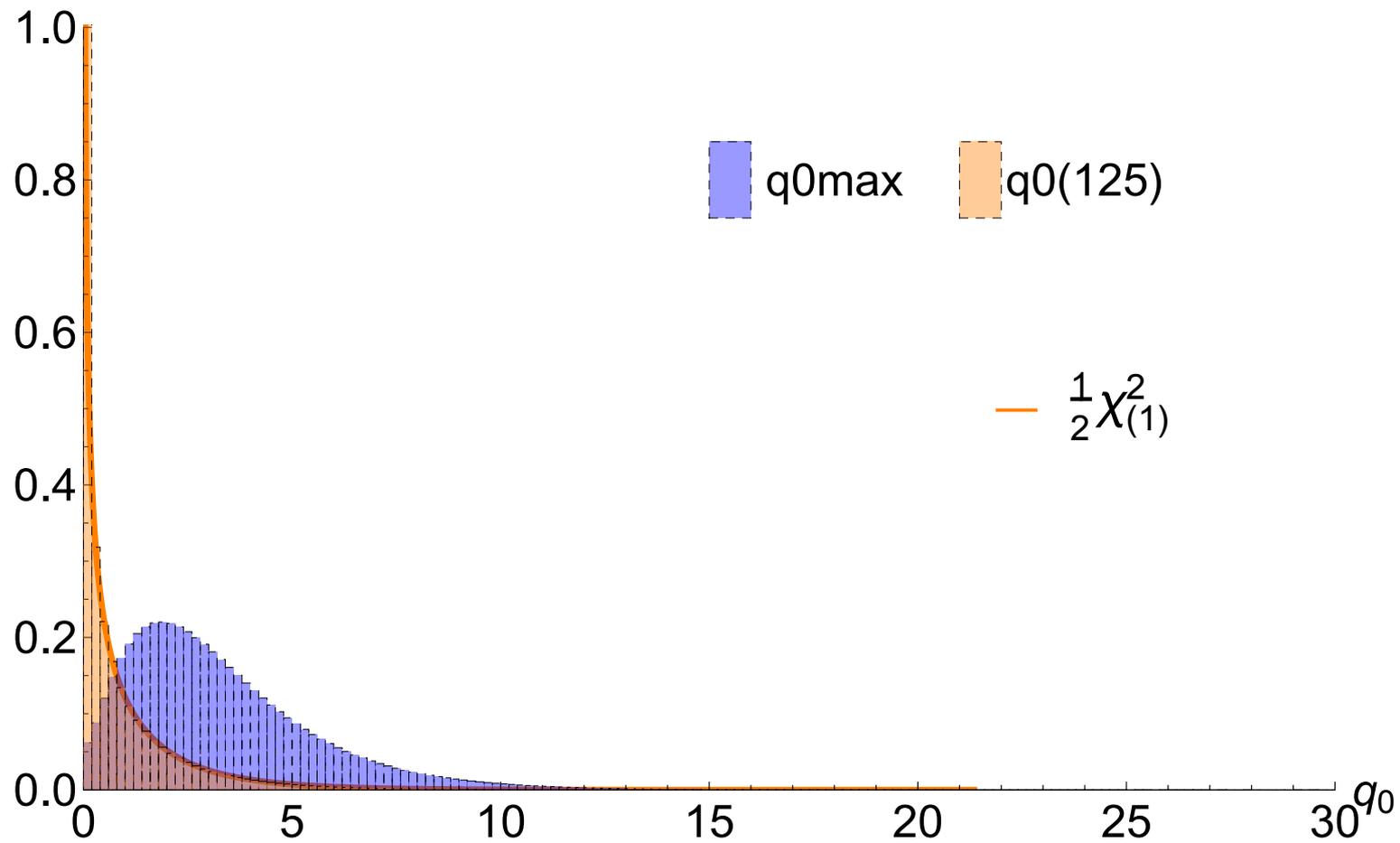


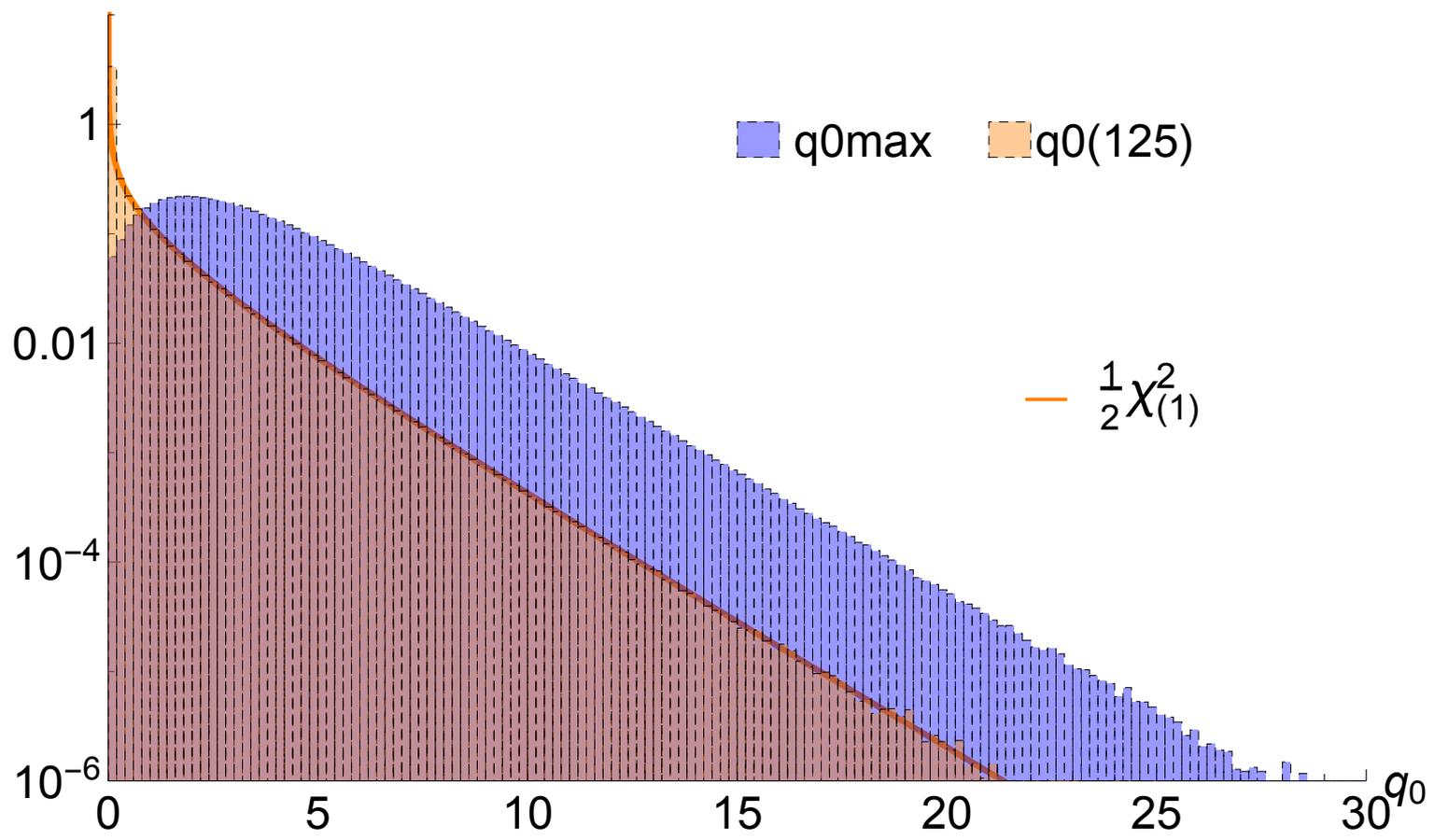
100 Million Experiments

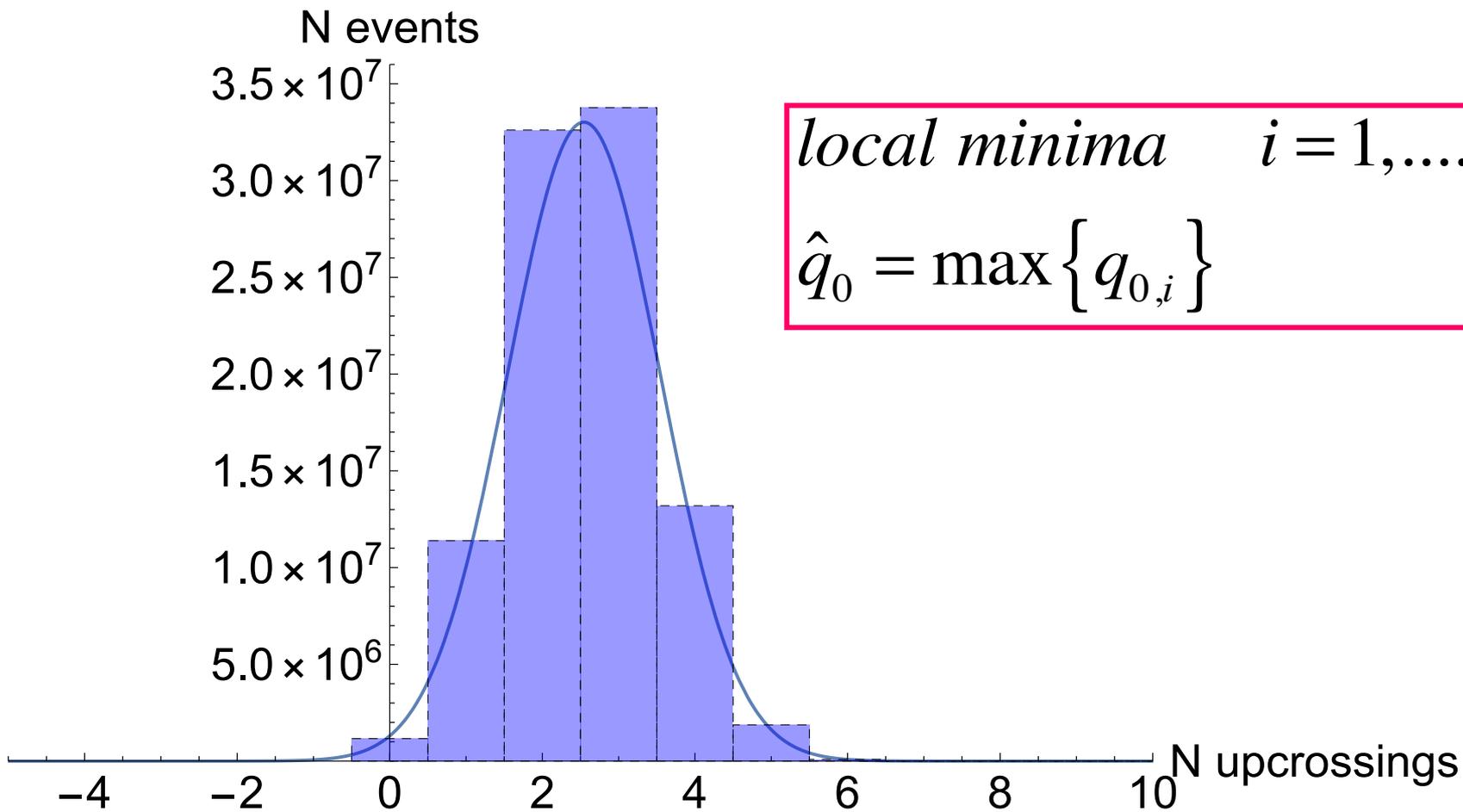


100 Million Experiments



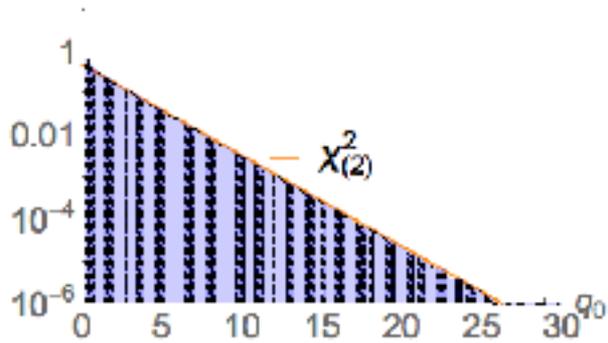




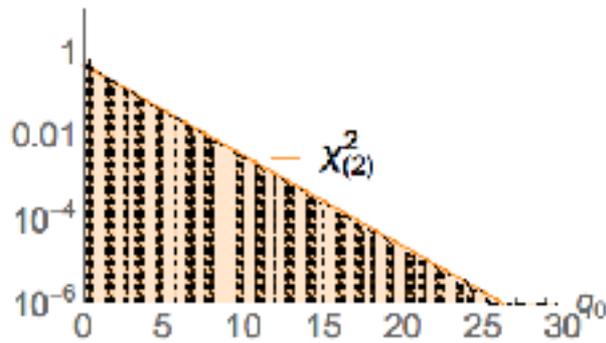


$$\forall i \quad q_{0,i} \sim \chi_2^2$$

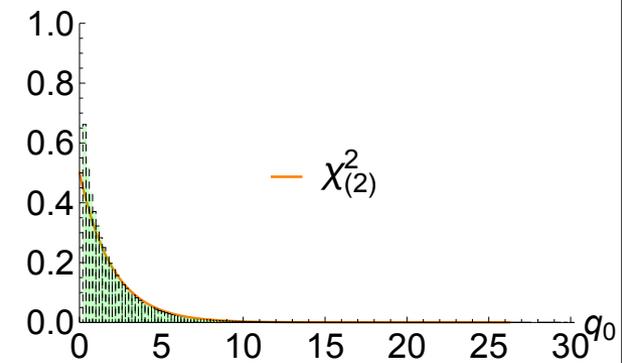
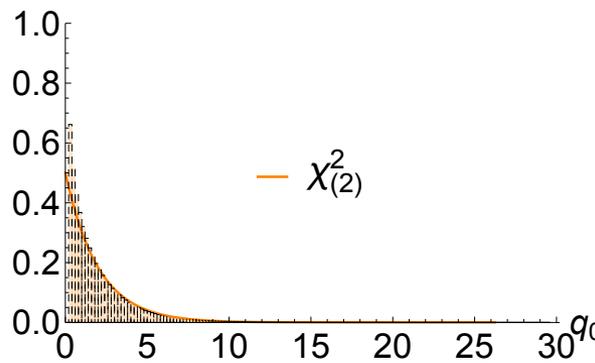
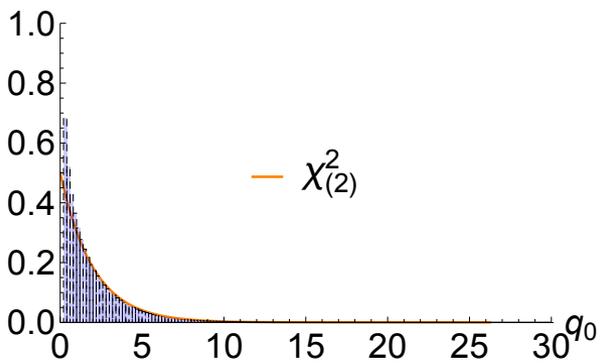
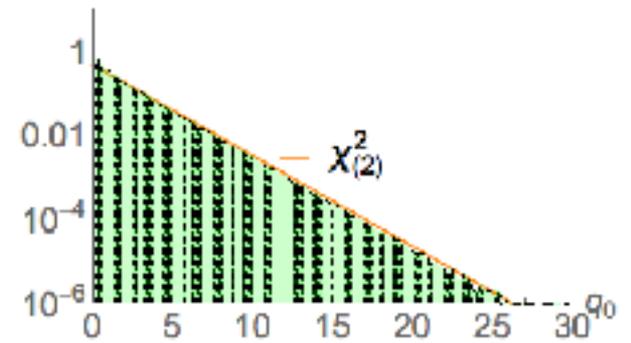
$q_{0,1}$

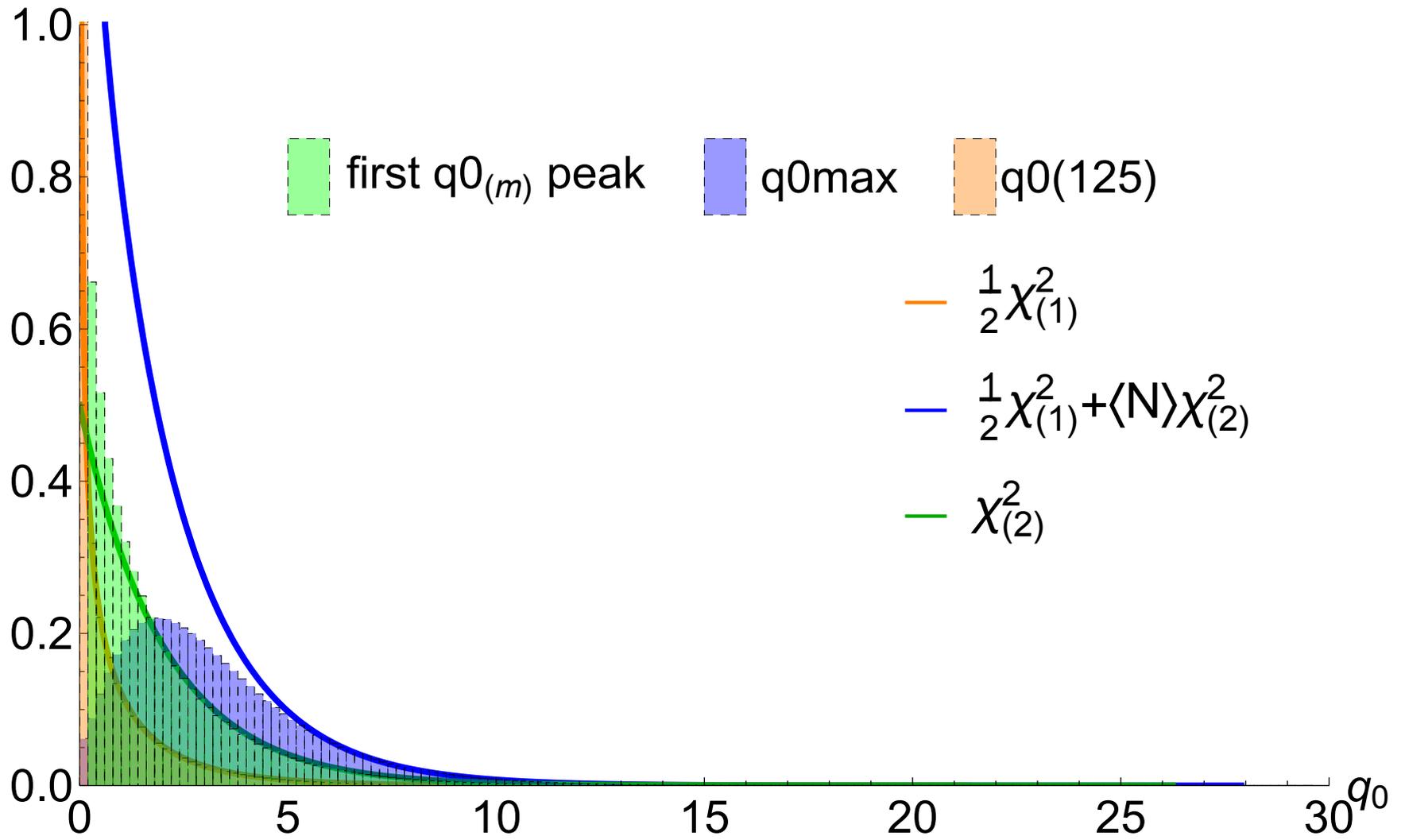


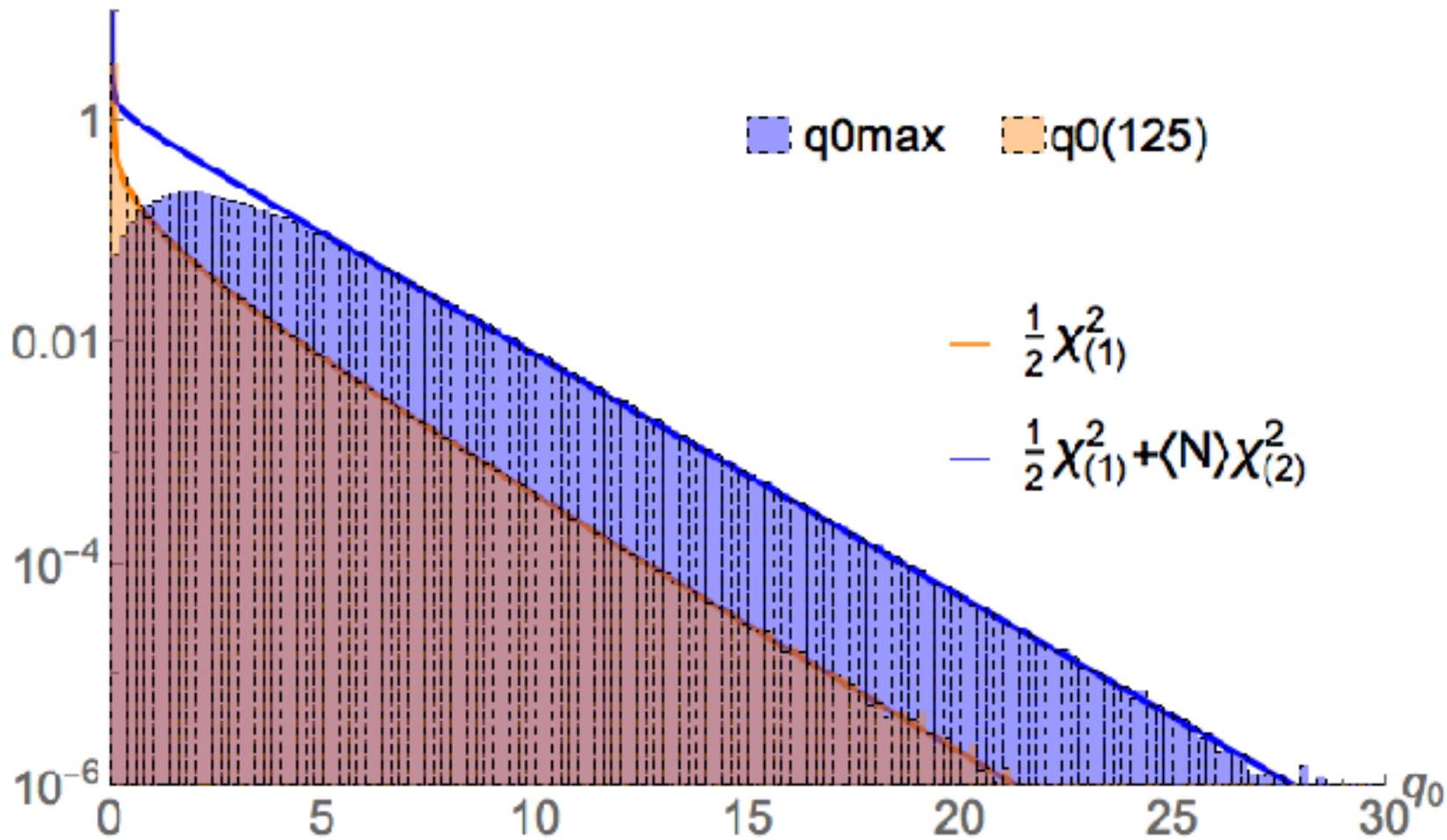
$q_{0,2}$



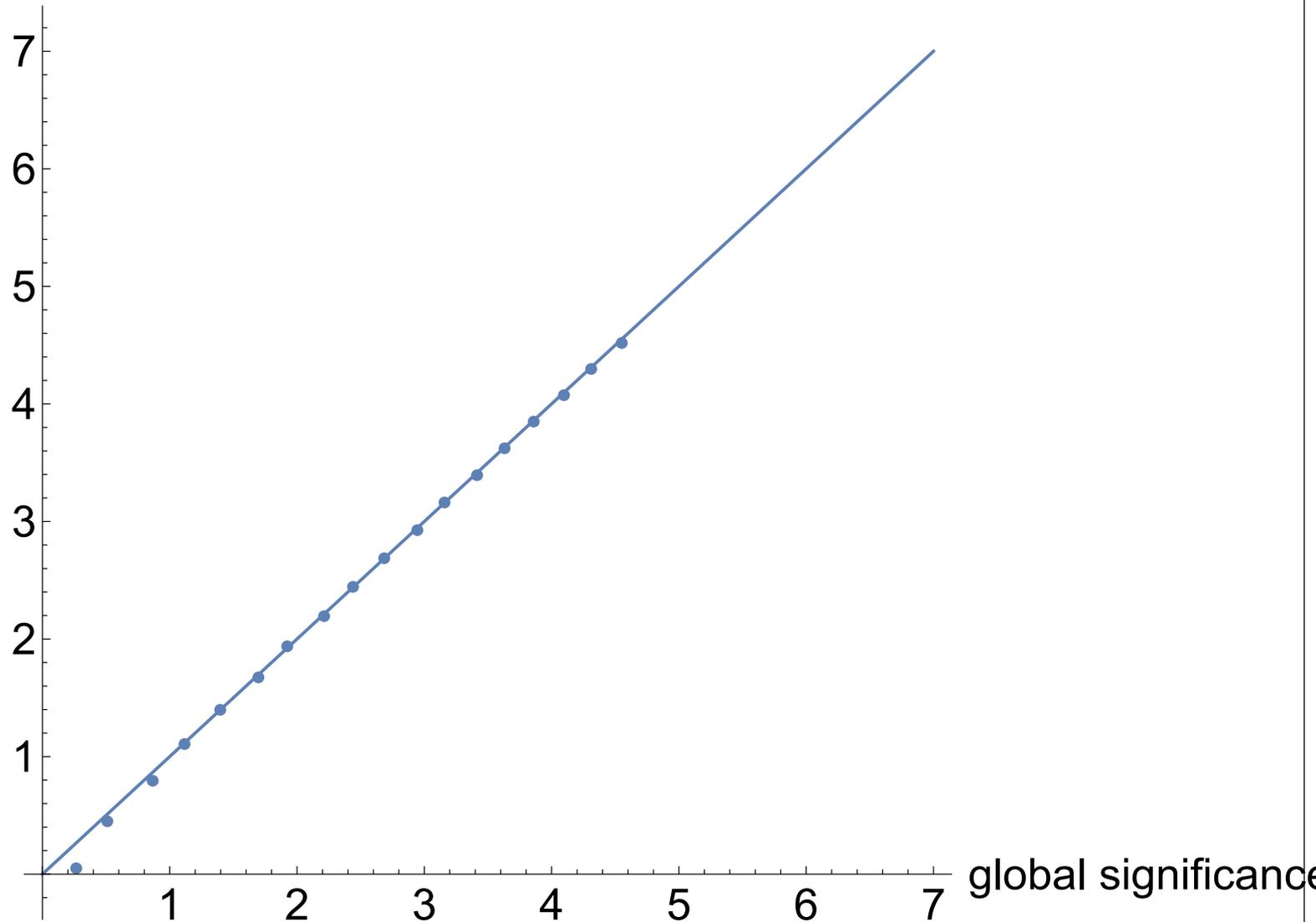
$q_{0,3}$







global significance formula



Trial Factor

40

30

20

10

0

— Formula
• Toys

0

1

2

3

4

5

6

7

Local Significance

$$trial \# \sim \sqrt{\frac{\pi}{2}} \mathcal{N} Z_{fix}$$



Why $\text{Trial\#} \sim Z_{\text{fix}}$?



Solution of the LEE problem

$$P(q(\hat{\theta}) > c) \sim P(\chi_1^2 > c) + \langle N(c_0) \rangle e^{-(c-c_0)/2}$$

$$\text{trial \#} \sim 1 + \sqrt{\frac{\pi}{2}} \mathcal{N} Z_{\text{fix}}$$

$$\langle N(c_0) \rangle = \mathcal{N} P(\chi_2^2 > c_0)$$

Where does the Z dependence come from?

View the results as if there are \mathcal{N} independent search regions

In each one there is a χ_2^2 distribution of $q_0(\mu, m)$

The mass is a dof even though it is undefined under the null

$$\sigma_{\hat{m}} \sim \frac{1}{Z} \Rightarrow \frac{\Delta m}{\sigma_{\hat{m}}} \sim Z$$

Why trial# ~ Z

$$\text{Var}(m) = \left[-E \left(\frac{\partial^2 \log \mathcal{L}}{\partial m^2} \right) \right]^{-1}$$

$$n \sim \text{Pois}(\mu s(m) + b) \approx e^{-(\mu s(m) + b)} (\mu s(m) + b)^n$$

$$\log \mathcal{L} = -\mu s(m) - b + n \log(\mu s(m) + b)$$

$$\frac{\partial \log \mathcal{L}}{\partial m} = -\mu \frac{\partial s(m)}{\partial m} + n \frac{\mu}{\mu s(m) + b} \frac{\partial s(m)}{\partial m}$$

$$\frac{\partial^2 \log \mathcal{L}}{\partial m^2} = -\mu \frac{\partial^2 s(m)}{\partial m^2} + n \frac{\mu}{\mu s(m) + b} \frac{\partial^2 s(m)}{\partial m^2} - n \frac{\mu^2}{(\mu s(m) + b)^2} \left(\frac{\partial s(m)}{\partial m} \right)^2$$

$$E[n] = \mu s(m) + b$$

$$E \left[\frac{\partial^2 \log \mathcal{L}}{\partial m^2} \right] = -\frac{\mu^2}{\mu s(m) + b} \left(\frac{\partial s(m)}{\partial m} \right)^2$$

$$\text{Var}[m] \sim \frac{1}{\mu} \sim \frac{1}{Z} \Rightarrow \sigma_{\hat{m}} \sim Z \Rightarrow \text{trial \#} \sim \frac{\text{range}}{\sigma_{\hat{m}}} \sim Z$$



A real life example

$$P(q_0 > u) \leq E[N_u] + P(q_0(0) > u)$$

$$E[N_u] = N_1 e^{-u/2}$$

$$N_1 \cong \langle N_{u_0} \rangle e^{u_0/2}$$

$$P(q_0 > u) = N_1 e^{-u/2} + \frac{1}{2} P(\chi_1^2 > u)$$

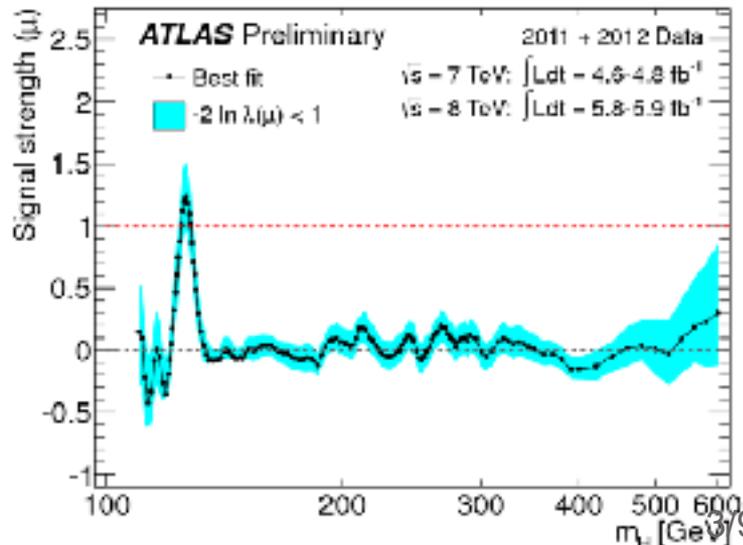
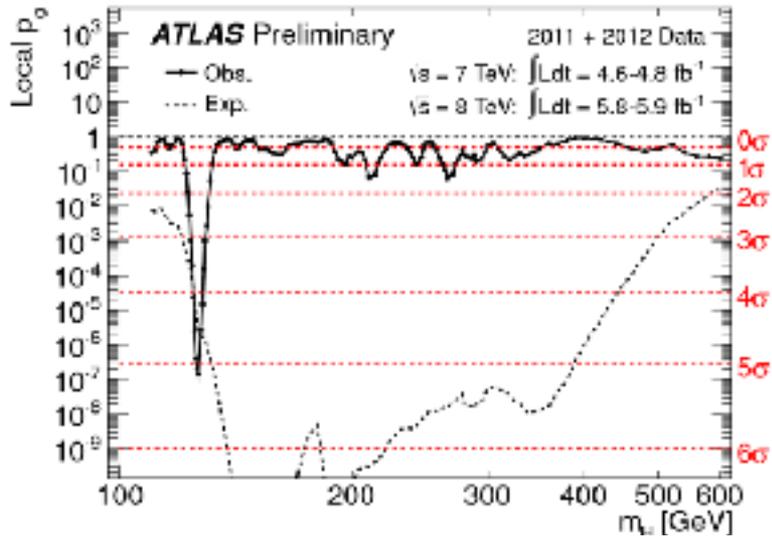
$$p_{global} = N_1 e^{-u/2} + p_{local}$$

$$p_{global} = \langle N_{u_0} \rangle e^{\frac{u_0 - u}{2}} + p_{local}$$

$$N_{u_0=0} = 9 \pm 3$$

$$p_{global} = 9 \cdot e^{-25/2} + O(10^{-7}) = 3.3 \cdot 10^{-5}$$

$$5\sigma \rightarrow 4\sigma \text{ trial}\# \sim 100$$



Example: The 750 GeV Resonance

Spin 0 2015

Largest significance

$m_x \sim 750 \text{ GeV}, \Gamma_x \sim 45 \text{ GeV}$ (6)

Local $Z = 3.9\sigma$

Any peak with $Z > 3.8\sigma$
with $m=500-2000$ will draw our attention

$$P_{global}(u) \approx p_{local}(u) + E(n_{u_0}) e^{-\frac{u_0 - u}{2}}$$

$$p_{local} = 5 \cdot 10^{-5}$$

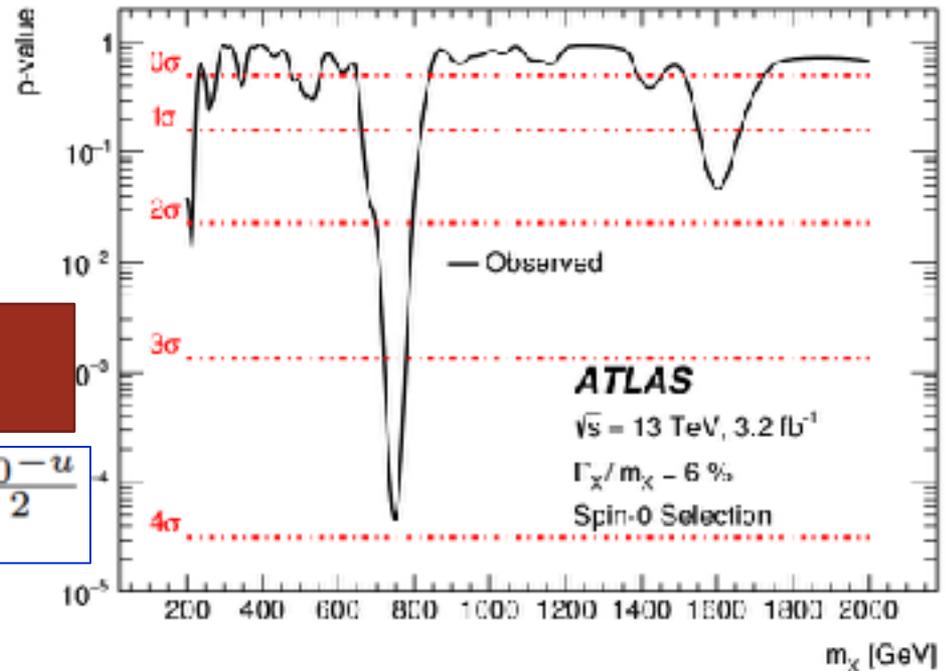
$$u_0 = 0$$

$$n_{u_0} = 7 \pm 2.6$$

$$u = Z^2 = 3.9^2 = 15.2$$

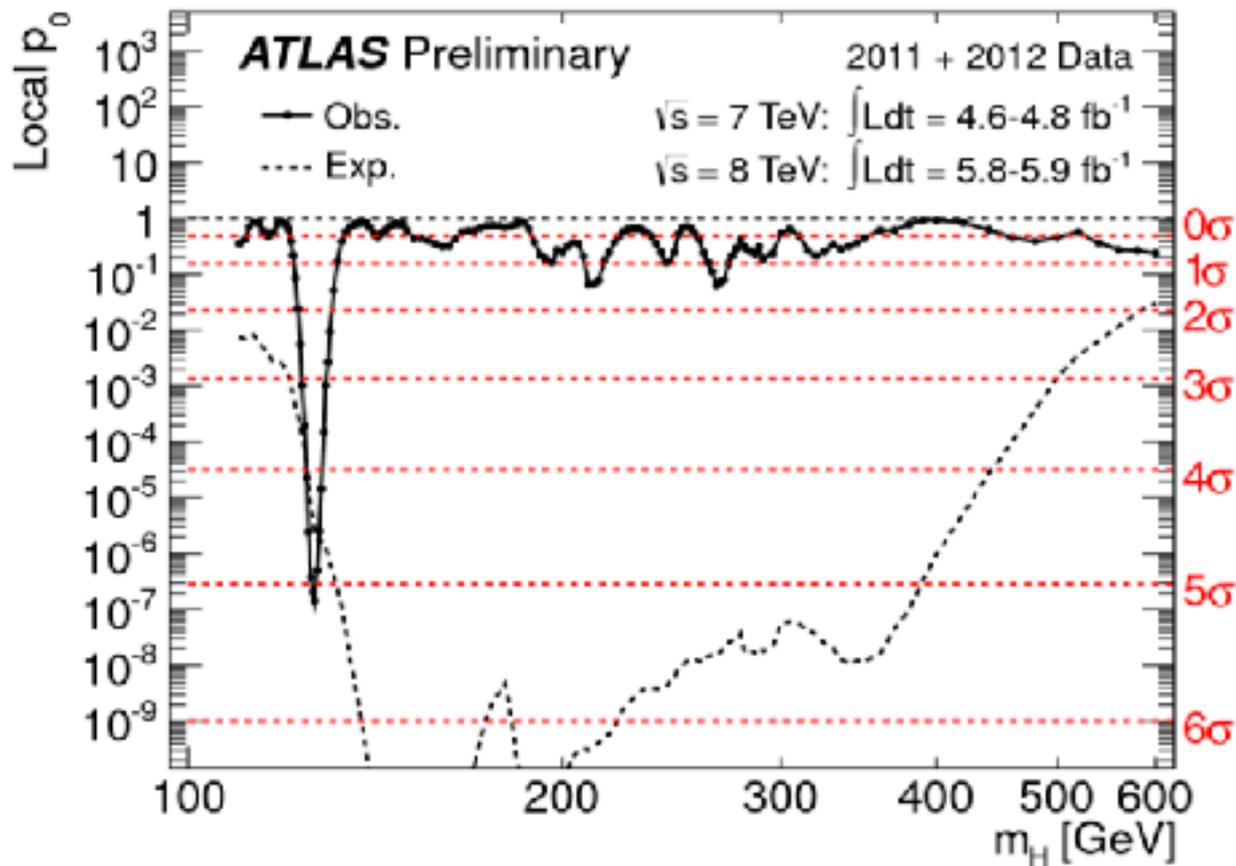
$$p_{global} = 5 \cdot 10^{-5} + (7 \pm 2.6) e^{-15.2/2} = (2.2 - 4.8) 10^{-3}$$

$$Z_{global} \sim 2.7 \pm 0.1\sigma$$



The LEE is even stronger when you consider another dimension (the width range (0-10%) should also be taken into account)

Here is a result of a p0 scan of the measured Higgs mass



1. Can you try to estimate the signal strength at the peak ($m_H=125$ GeV)
2. What is p_0 value when $Z=0$ sigma?
3. What is the number of upcrossings at $u=0$ (with the error)?

$$p_{global} = \left\langle N_{\nu_0} \right\rangle e^{\frac{\nu_0 - u}{2}} + p_{local}$$

4. remember the formulae:
what is p_{local} ?
5. Find the global p-value
6. What is the corresponding significance taking the Look Elsewhere Effect into account?
7. Calculate the trial factor



The 2D LEE



Define the Problem

- Let $n = \mu s(m, \Gamma) + b$
- m, Γ are nuisance parameters undefined under the null hypothesis $\mu = 0$
- What is the pdf of

$$\hat{q}_0 \equiv q_0(\hat{m}, \hat{\Gamma}) = -2 \ln \frac{L(\mu = 0)}{L(\hat{\mu}, \hat{m}, \hat{\Gamma})} = \max_{m, \Gamma} q_0(m, \Gamma)$$

under the null hypothesis

Define the Problem

- To generalize the problem, let Θ be the nuisance parameter, undefined under the null hypothesis, and let us try to find out the pdf of

$$\hat{q}_0 \equiv q_0(\hat{\theta}) = -2 \ln \frac{L(\mu=0)}{L(\hat{\mu}, \hat{\theta})} = \max_{\theta} q_0(\theta)$$

for which we want to calculate

$$p\text{-value} = P\left(\max_{\theta} [q_0(\theta)] \geq u\right), \quad u = Z^2$$

Chi Squared Random Field

- For fixed θ $q_0(\theta) = -2 \ln \frac{L(\mu=0)}{L(\hat{\mu}, \theta)} \sim \chi_1^2$

- $q_0(\theta)$ is a chi squared random field over the space of θ

(a random variable indexed by a continuous parameter(s))

- We are interested in

$$\hat{q}_0 \equiv q_0(\hat{\theta}) = -2 \ln \frac{L(\mu=0)}{L(\hat{\mu}, \hat{\theta})} = \max_{\theta} q_0(\theta)$$

for which we want to calculate

$$p\text{-value} = P\left(\max_{\theta} [q_0(\theta)] \geq u \right), \quad u = Z^2$$



Chi Squared Random Field

- We are only interested in positive signals (downward fluctuations of the background are not considered as an evidence against the background)

$$q_0(\theta) = \begin{cases} -2 \log \frac{\mathcal{L}(\mu = 0)}{\mathcal{L}(\hat{\mu}, \theta)} & q_0(\theta) \sim \frac{1}{2} \chi_1^2 \\ 0 & \end{cases}$$

[H. Chernoff, Ann. Math. Stat. 25, 573578 (1954)]



Chi Squared Random Field

- We are only interested in positive signals
(downward fluctuations of the background are not considered as an evidence against the background)

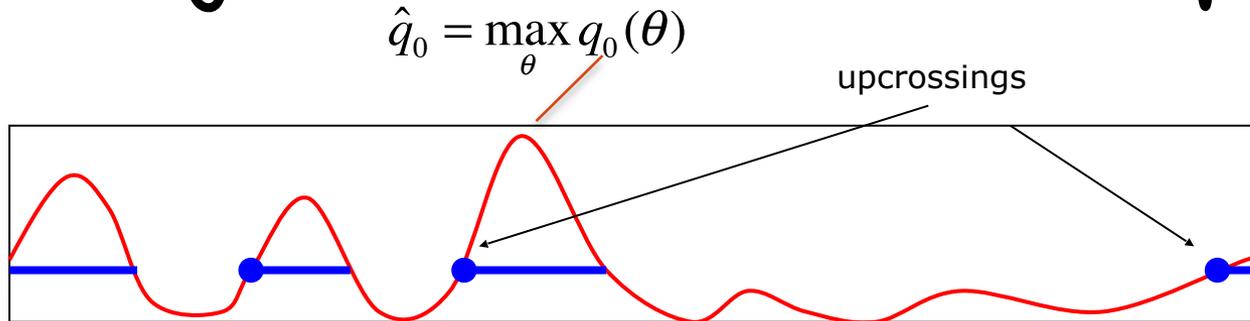
$$q_0(\theta) = \begin{cases} -2 \log \frac{L(\mu = 0)}{L(\hat{\mu}, \theta)} & q_0(\theta) \sim \frac{1}{2} \chi_1^2 \\ 0 & \end{cases}$$

[H. Chernoff, Ann. Math. Stat. 25, 573578 (1954)]

- $q_0(\theta) = \left(\frac{\hat{\mu}(\theta)}{\sigma} \right)^2$ $\hat{\mu}(\theta)$ is a Gaussian Random Field over θ

1-D Random Fields

- In 1-D points where the field becomes larger than u are called upcrossings.



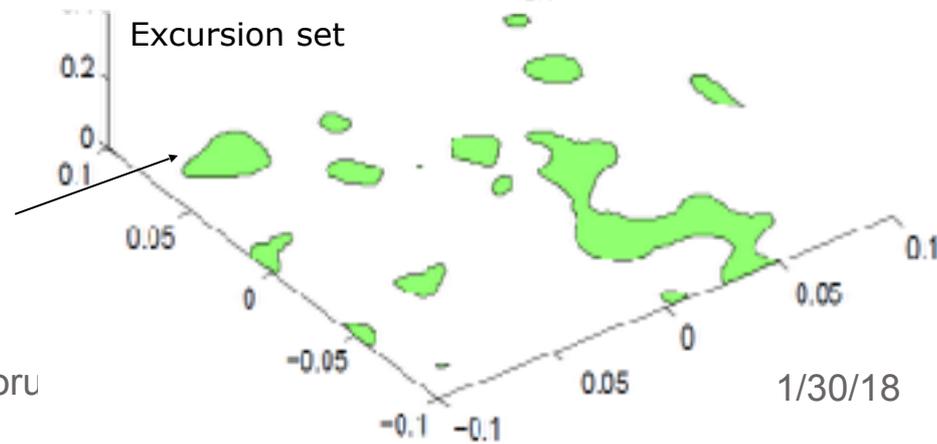
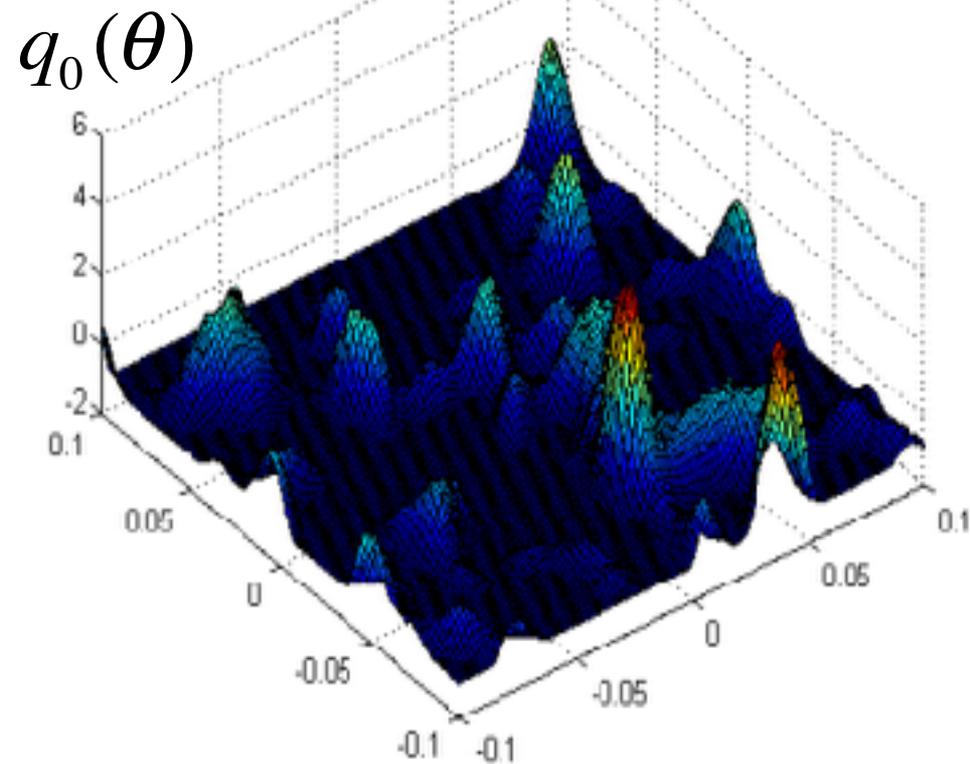
- The probability that the global maximum is above the level u is called **exceedance probability**.

(p-value of $q_0(\hat{\theta})$) $p = P\left(\max_{\theta} [q_0(\theta)] \geq u\right), u = Z^2$



Random fields (>1 D)

- The set of points where the field has values larger than some number u is called the excursion set A_u above the level u .



Random fields

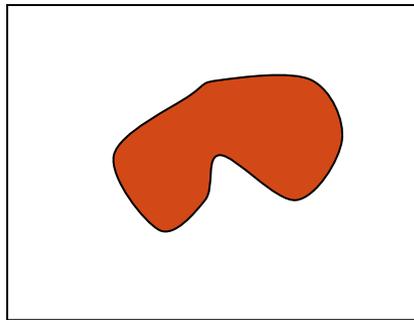
- Fortunately, quite a lot of statistical literature on the properties of random fields in D - dimensions
- Applications in Cosmology, Brain mapping, Oceanography ...

- [3] R.J. Adler and A.M. Hasofer, *Level Crossings for Random Fields*, Ann. Probab. 4, Number 1 (1976), 1-12.
- [4] R.J. Adler, *The Geometry of Random Fields*, New York (1981), Wiley, ISBN: 0471278440.
- [5] K.J. Worsley, S. Marrett, P. Neelin, A.C. Vandal, K.J. Friston and A.C. Evans, *A Unified Statistical Approach for Determining Significant Signals in Location and Scale Space Images of Cerebral Activation*, Human Brain Mapping 4 (1996) 58-73.
- [6] R.J. Adler and J.E. Taylor, *Random Fields and Geometry*, Springer Monographs in Mathematics (2007). ISBN: 978-0-387-48112-8.
- [9] J. Taylor, A. Takemura and R.J. Adler, *Validity of the expected Euler characteristic heuristic*, Ann. Probab. 33 (2005) 1362-1396.

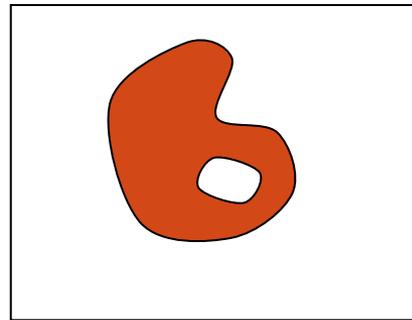


Euler characteristic

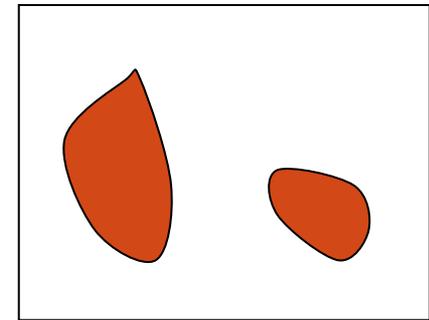
- Number of disconnected components minus number of 'holes'



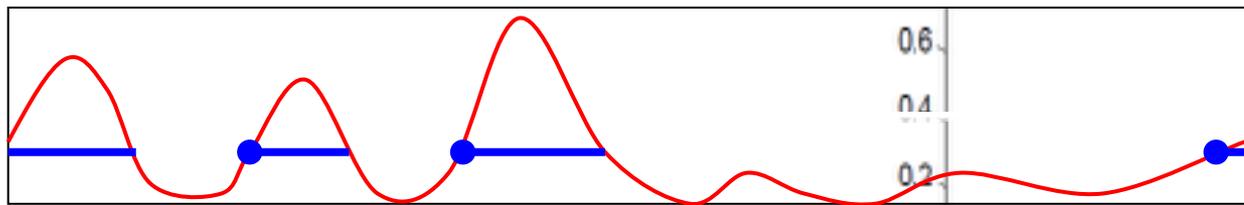
$\varphi=1$



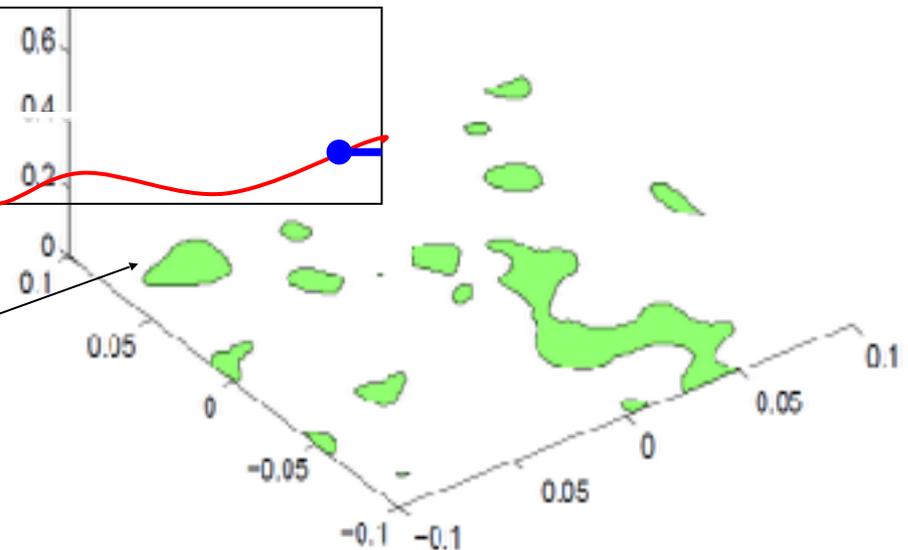
$\varphi=0$



$\varphi=2$



Excursion set



The n-dimensional case

[R.J. Adler and J.E. Taylor, *Random Fields and Geometry* (2007), Springer Monographs in Mathematics]

- The upcrossings formula is a special case of a more general result which gives the expectation of the Euler characteristic of the excursion set of a random field over a general n-dimensional manifold

$$E[\varphi(A_u)] = \sum_{d=0}^D \mathcal{N}_d \rho_d^s(u)$$

A_u is the excursion set of the field above a level u
(set of points where $q_o(\theta) > u$)

$\varphi(A_u)$ is its Euler characteristic

ρ_d are 'universal' functions

(depend only on the level u and s , number of poi)



Adler et. al. Formula

$$E[\vartheta(A_u)] = \sum_{d=0}^D \mathcal{N}_d \rho_d^s(u)$$

n is the Dimension
(number of Nuisance parameters
undefined under the null hypothesis)

- For a Chi Squared field with S parameters of interest and dimension D

$D = 1, s \text{ poi}$

$$\rho_0(u) = P(\chi_1^2 > u)$$

$$\rho_1(u) = u^{(s-1)/2} e^{-u/2}$$

–

$D = 2, s \text{ poi}$

$$\rho_0(u) = P(\chi_2^2 > u)$$

$$\rho_1(u) = u^{(s-1)/2} e^{-u/2}$$

$$\rho_2(u) = u^{(s-2)/2} (u - (s-1)) e^{-u/2}$$

$D = 2, s = 1$

$$\rho_0(u) = P(\chi_2^2 > u)$$

$$\rho_1(u) = e^{-u/2}$$

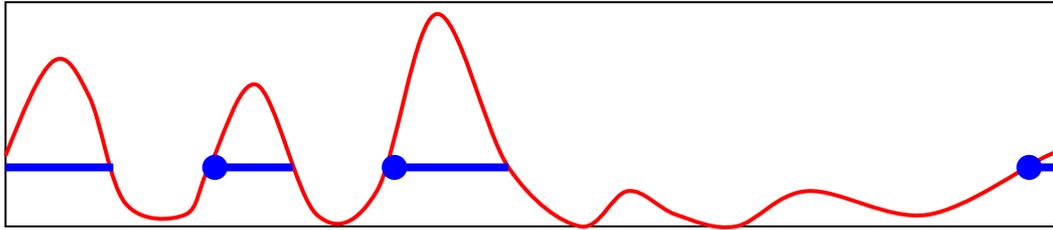
$$\rho_2(u) = \sqrt{u} e^{-u/2}$$

$$1D: E[\vartheta(A_u)] = \frac{1}{2} P(\chi_1^2 > u) + \mathcal{N}_1 e^{-u/2}$$

$$2D: E[\vartheta(A_u)] = \frac{1}{2} P(\chi_2^2 > u) + (\mathcal{N}_1 + \mathcal{N}_2 \sqrt{u}) e^{-u/2}$$



1D Euler characteristic



$$E[\varphi(A_u)] = \sum_{d=0}^n N_d \rho_d(u)$$

In 1 dimension:

$$\varphi(A_u) = N_u + 1_{[q_0(0) > u]}$$

$$\begin{aligned} E[\varphi(A_u)] &= E[N_u] + P(q_0(0) > u) \\ &= N_0 P(\chi_1^2 > u) + N_1 e^{-u/2} \end{aligned}$$

$$N_0 = \varphi(\text{manifold}) = 1$$

$$E[\varphi(A_u)] = P(\chi_1^2 > u) + N_1 e^{-u/2}$$

This is Davies Formula

In general for high-level excursions

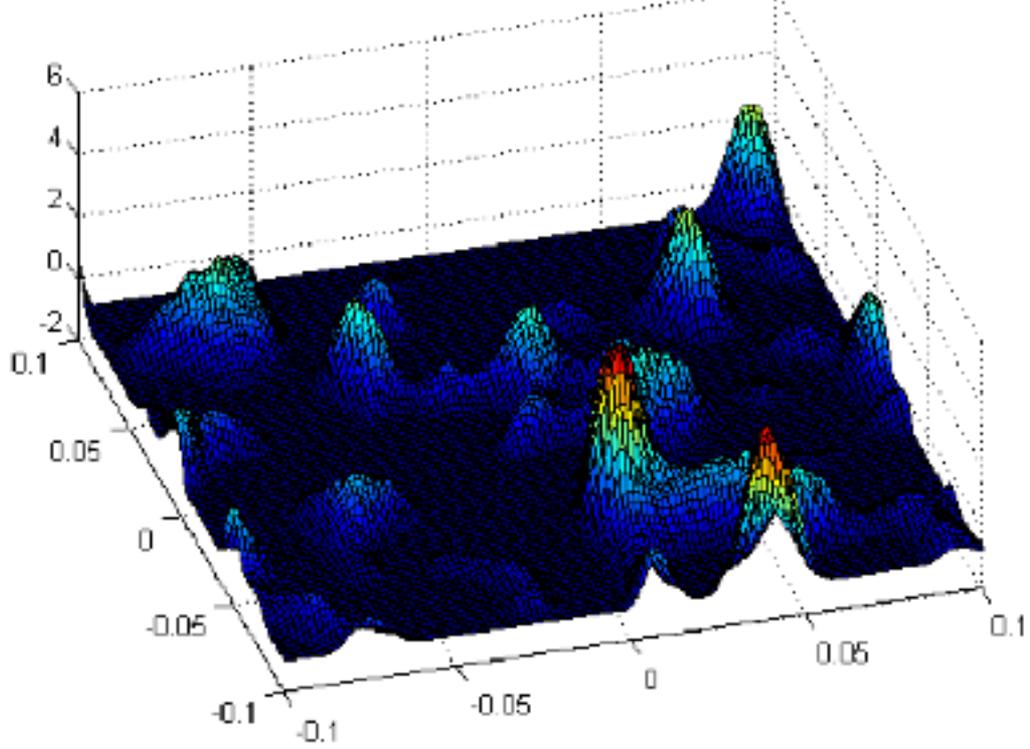
The general case

$$N_0 = \varphi(\text{manifold})$$

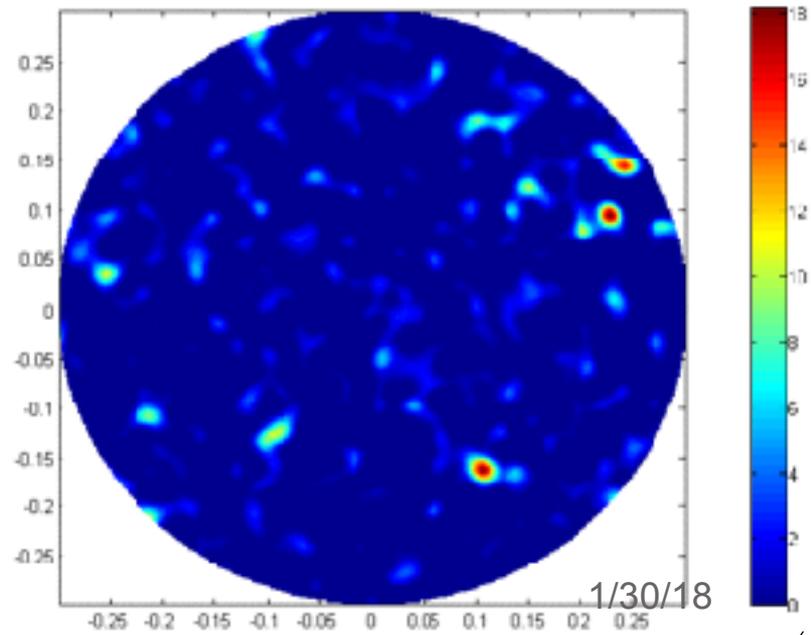
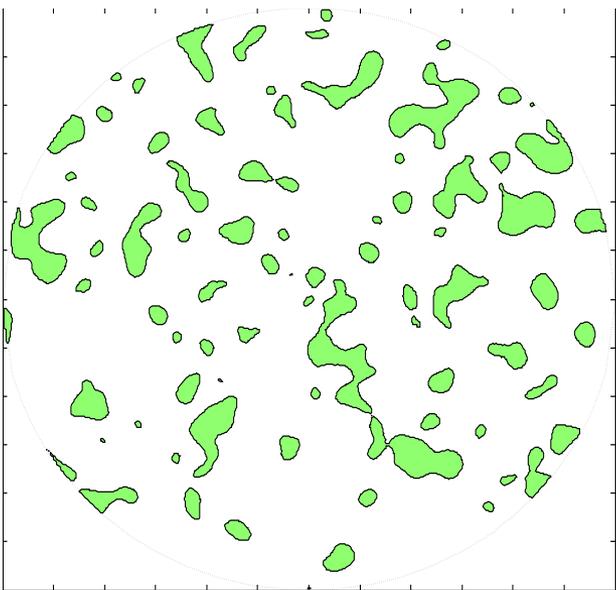
$$\rho_0(u) = P(\chi_s^2 > u)$$

$$E[\varphi(A_u)] \xrightarrow{u \gg 1} P\left(\max_{\theta} [q_0(\theta)] \geq u\right)$$





Excursion set
($u=1$)

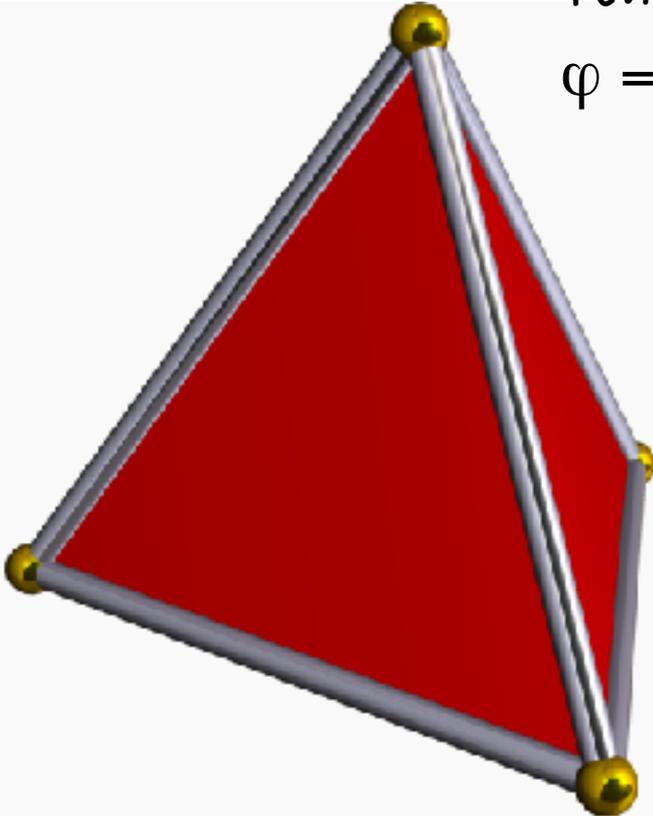


Calculation of the Euler characteristic

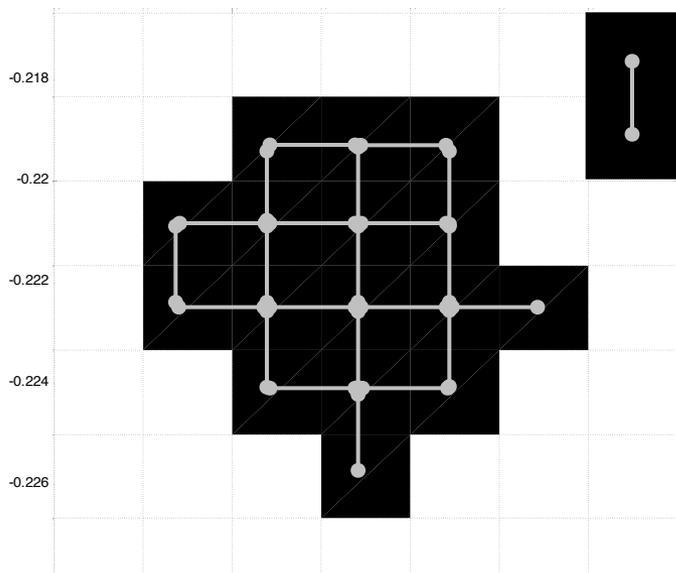
Tetrahedron

$$\varphi = V - E + F = \#vertices - \#edges + \#faces$$

$$\varphi = 4 - 6 + 4 = 2$$



Calculation of the Euler characteristic

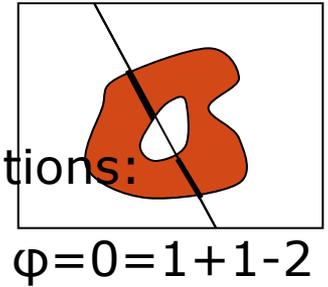


- Usually we have $q(\theta)$ calculated on a grid of points
- Calculation of the E.C. is straightforward:
- $\varphi = \# \text{vertices} - \# \text{edges} + \# \text{faces}$
- Generalizes to higher dimensions

$$\varphi = 18(\text{points}) - 23(\text{edges}) + 7(\text{faces}) = 2$$

Slicing

- Exploit the azimuthal angle symmetry to reduce computations:

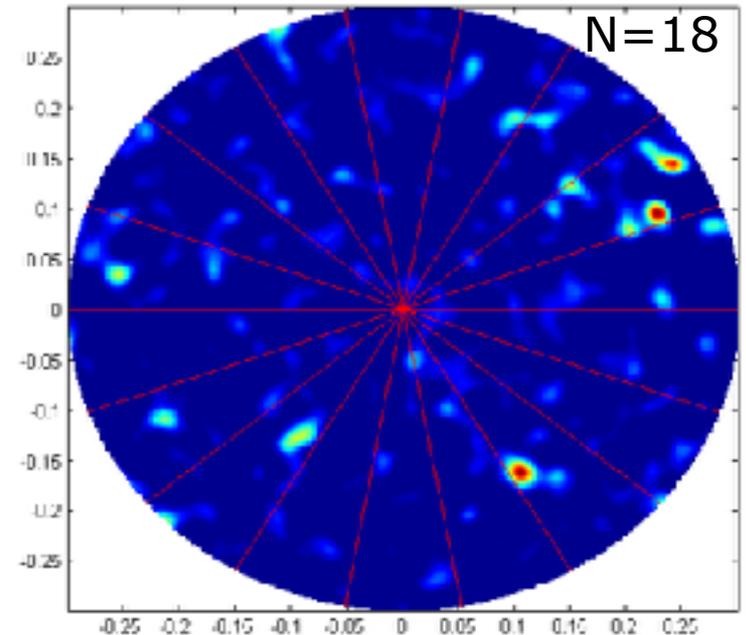
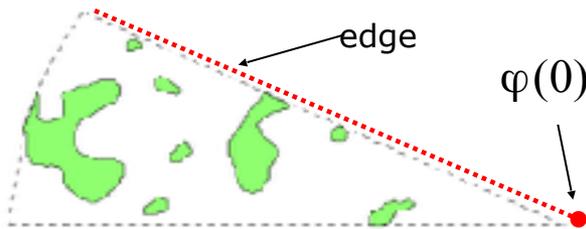


$$\varphi(A \cup B) = \varphi(A) + \varphi(B) - \varphi(A \cap B)$$

Divide to N slices:

$$\varphi = \sum_i [\varphi(\text{slice}_i) - \varphi(\text{edge}_i)] + \varphi(0)$$

$$E[\varphi] = N \times (E[\varphi(\text{slice})] - E[\varphi(\text{edge})]) + \varphi(0)$$



2-d example: search for neutrino sources (IceCube)

For a χ^2 field in 2 dimensions:

$$E[\vartheta(A_u)] = \frac{1}{2} P(\chi^2_2 > u) + (\mathcal{N}_1 + \mathcal{N}_2 \sqrt{u}) e^{-u/2}$$

Estimate $E[\varphi]$ at two levels, e.g. 0 and 1, and solve for \mathcal{N}_1 and \mathcal{N}_2

From 20 bkg. Simulations:

$$\langle \varphi_0 \rangle = 33.5 \pm 2$$

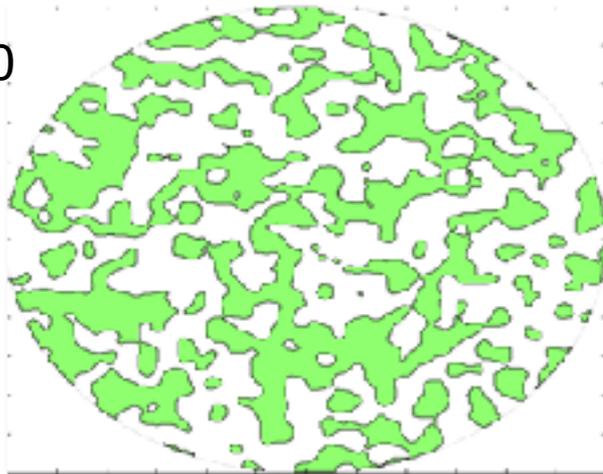
$$\langle \varphi_1 \rangle = 94.6 \pm 1.3$$

↓

$$\mathcal{N}_1 = 33 \pm 2$$

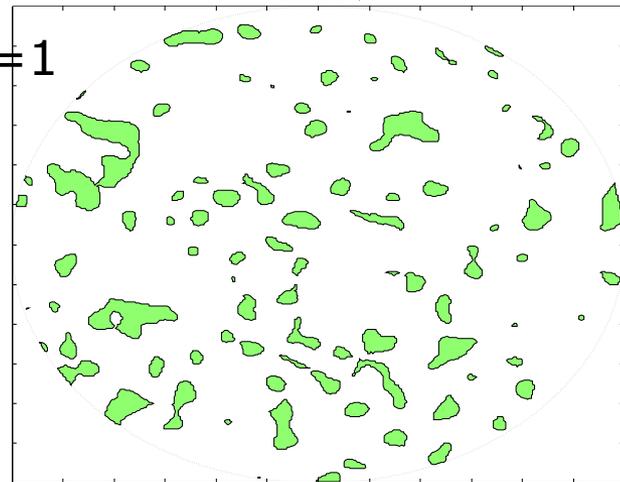
$$\mathcal{N}_2 = 123 \pm 3$$

$u=0$



$\varphi=35$

$u=1$



$\varphi=95$

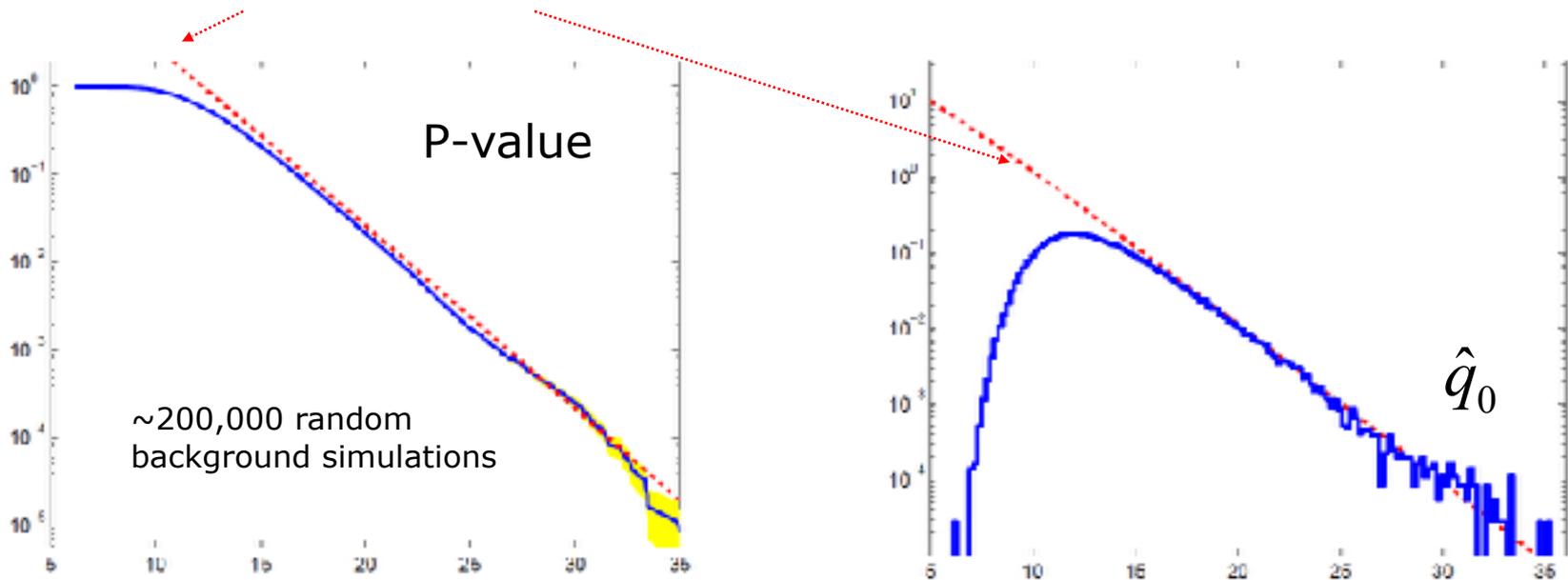


2-d example: search for neutrino sources (IceCube)

$$E[\vartheta(A_u)] = \frac{1}{2} P(\chi_2^2 > u) + (\mathcal{N}_1 + \mathcal{N}_2 \sqrt{u}) e^{-u/2}$$

$$\mathcal{N}_1 = 33 \pm 2$$

$$\mathcal{N}_2 = 123 \pm 3$$



e.g.: $P(\max q_0 > 30) = (2.5 \pm 0.4) \times 10^{-4}$ (estimated)

E.C. Formula : $(2.28 \pm 0.06) \times 10^{-4}$

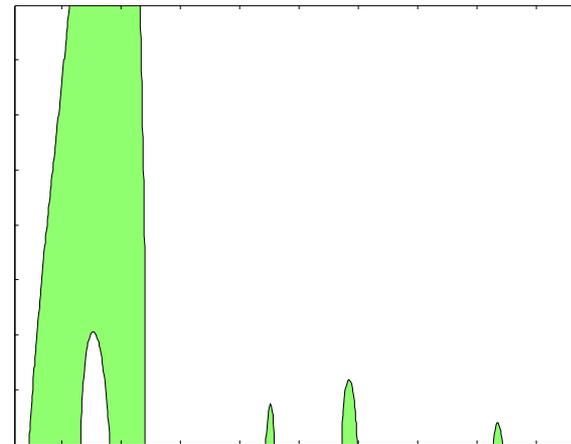
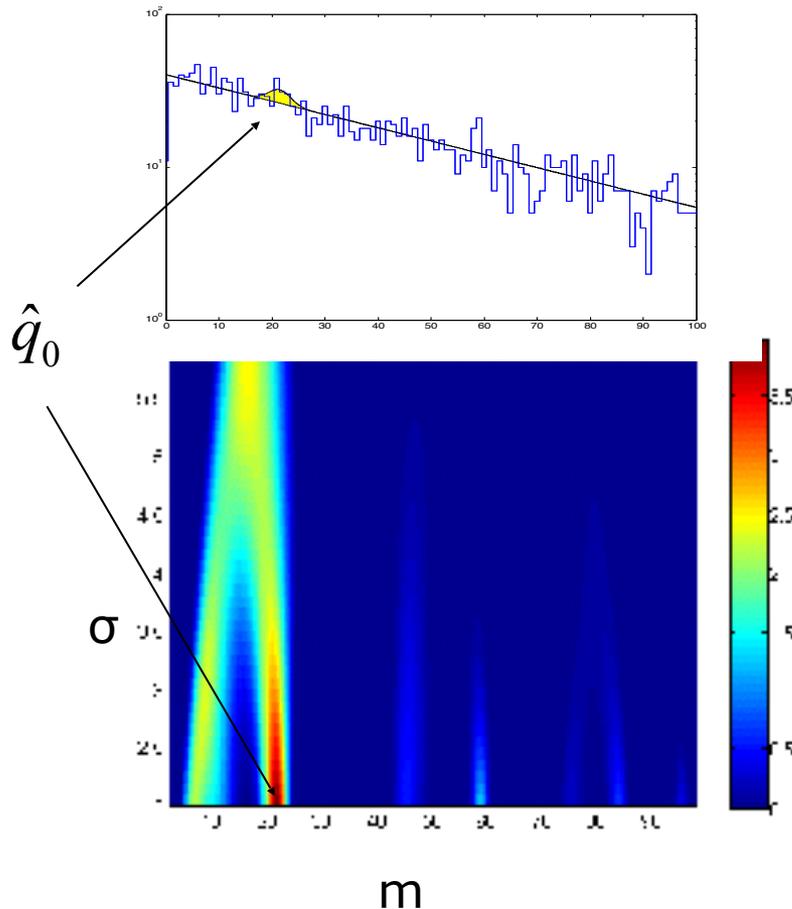


2-D exapmle #2: resonance search with unknown width

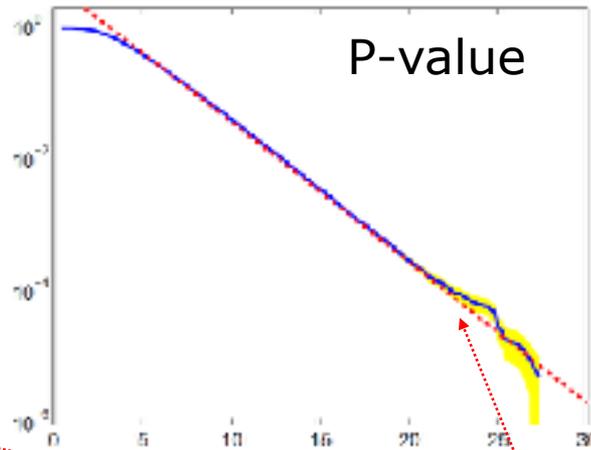
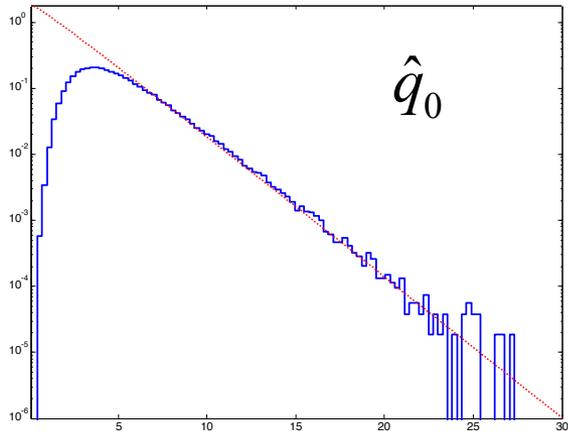
- Gaussian signal on exponential background
- Toy model : $0 < m < 100$, $2 < \sigma < 6$
- Unbinned likelihood:

$$\mathcal{L} = \prod_i \frac{N_s f_s(x_i) + N_b f_b(x_i)}{N_s + N_b} \times \text{Pois}(N | N_s + N_b)$$

$$f_s(x; m, \sigma) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-m)^2}{2\sigma^2}} \quad f_b(x) = ce^{-cx}$$



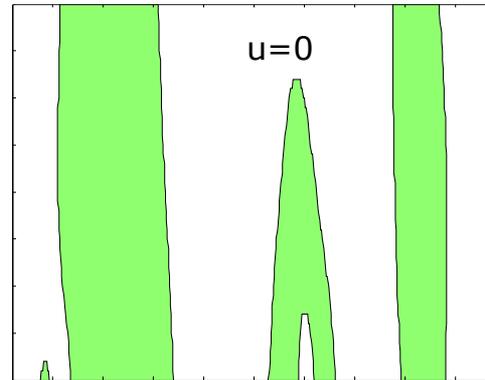
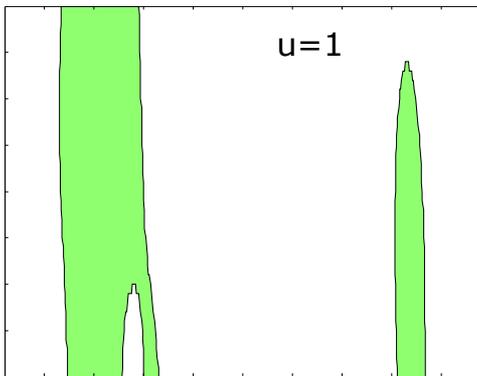
2-D example #2: resonance search with unknown width



Excellent approximation above the $\sim 2\sigma$ level

$$\langle \varphi_1 \rangle = 3 \pm 0.16$$

$$\langle \varphi_0 \rangle = 4.5 \pm 0.2$$



$$E[\vartheta(A_u)] = \frac{1}{2} P(\chi_2^2 > u) + (\mathcal{N}_1 + \mathcal{N}_2 \sqrt{u}) e^{-u/2}$$

$$\mathcal{N}_1 = 4 \pm 0.2$$

$$\mathcal{N}_2 = 0.7 \pm 0.3$$



2015

2D Scan

Largest significance

$m_x \sim 750\text{GeV}, \Gamma_x \sim 45\text{GeV}(6\%)$

Local $Z = 3.9\sigma$

$m=200-2000\text{ GeV}$
 $\Gamma_x/m_x=0-10\%$

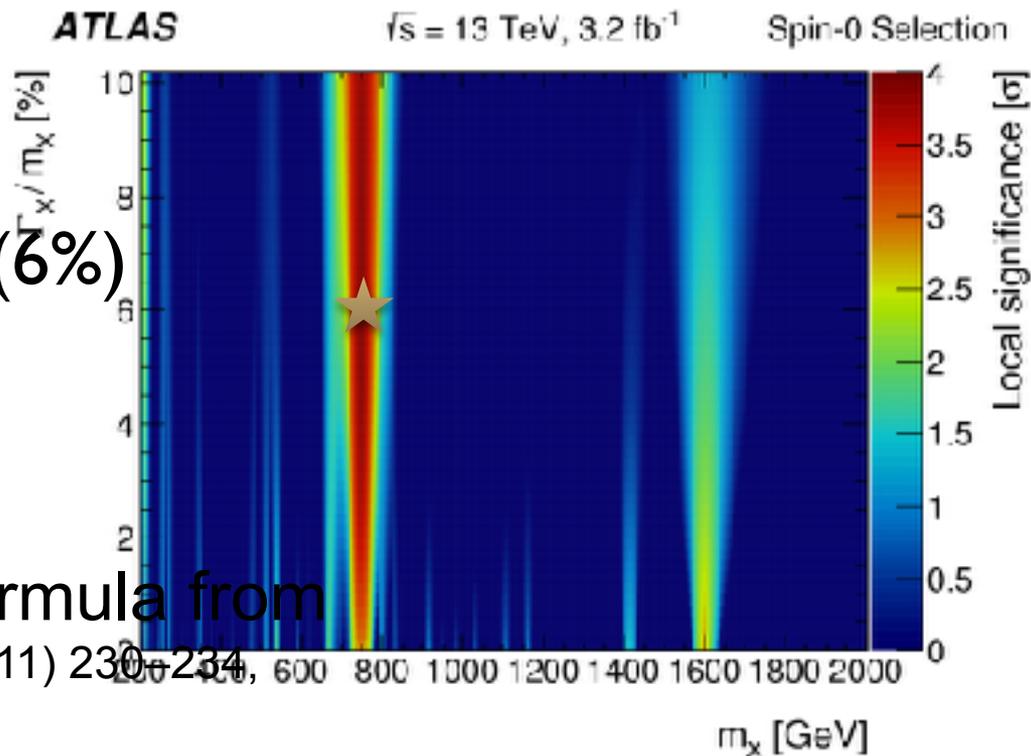
Use toys or asymptotic formula from

O. Vitells et. al. Astropart. Phys. 35 (2011) 230-234,
 arXiv:1105.4355

$$Z_{local} = 3.9\sigma$$

$$Z_{global} = 2.1\sigma$$

2.1 σ is not something to write home about



Summary

$$P_{global}(s=1, D=1) \approx E[\vartheta(A_u)] = \frac{1}{2} P(\chi_1^2 > u) + \mathcal{N}_1 e^{-u/2}$$

$$P_{global}(s=1, D=2) \approx E[\vartheta(A_u)] = \frac{2}{2} P(\chi_2^2 > u) + (\mathcal{N}_1 + \mathcal{N}_2 \sqrt{u}) e^{-u/2}$$

- The procedure for estimating the p-value is simple and reliable.
- The Euler characteristic formula provides a practical way of estimating the look-elsewhere effect.
- It is easily expandable to s p.o.i and D NPs (undefined under the null hypothesis)



End of Lectures
Thank You

Eilam Gross

