

Eilam Gross

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

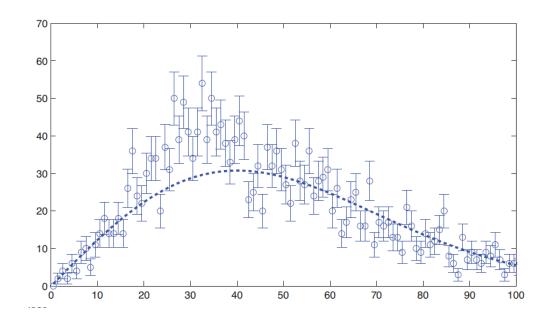
Eur. Phys. J. C 70 (2010) 525

O. Vitells and E. G., Estimating the significance of a signal in a multidimensional search,

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

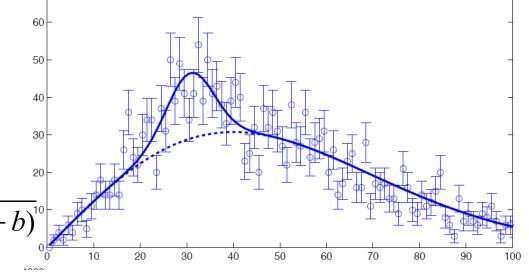
Search for resonances in diphoton events at s√=13 TeV with the ATLAS detector, ATLAS collaboration, arXiv 1606.03833

•Is there a signal here?



- •Looks like @ m=30
- •What is its significance?
- What is your test statistic?

$$q_{fix,obs} = -2 \ln \frac{L(b)}{L(\hat{\mu}s(m=30)+b)^{10}}$$



•Test statistic

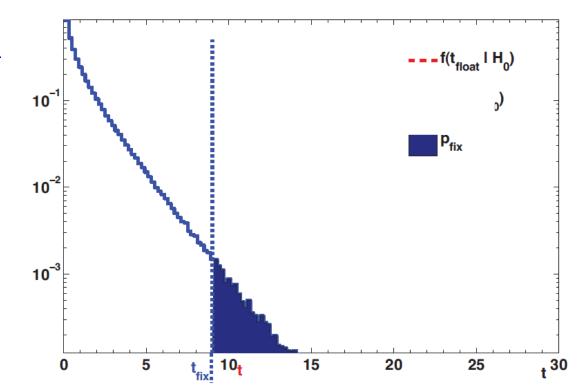
$$q_{fix,obs} = -2 \ln \frac{L(b)}{L(\hat{\mu}s(m=30)+b)}$$

- •What is the p-value?
- generate the PDF

$$f(q_{fix} | H_0)$$

and find the **p-value**Wilks theorem:

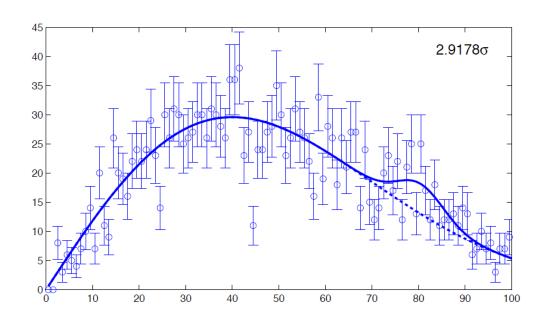
$$f(q_{fix} \mid H_0) \sim \chi_1^2$$



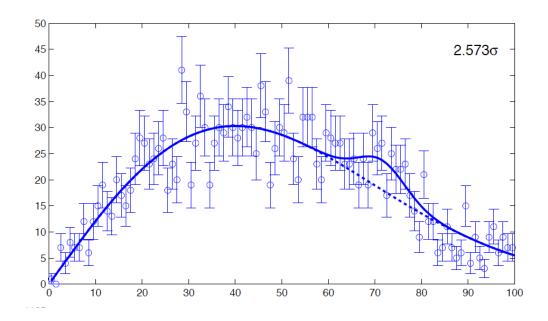
$$p_{fix} = \int_{q_{fix},obs}^{\infty} f(q_{fix} \mid H_0) dq_{fix}$$



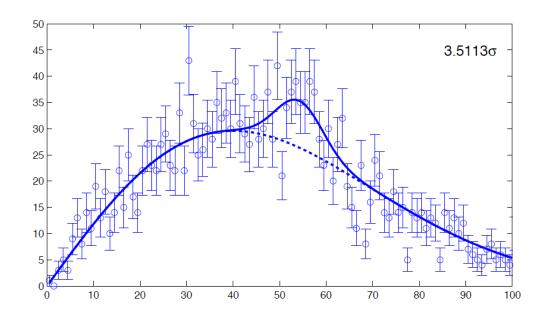
•Would you ignore this signal, had you seen it?



•Or this?



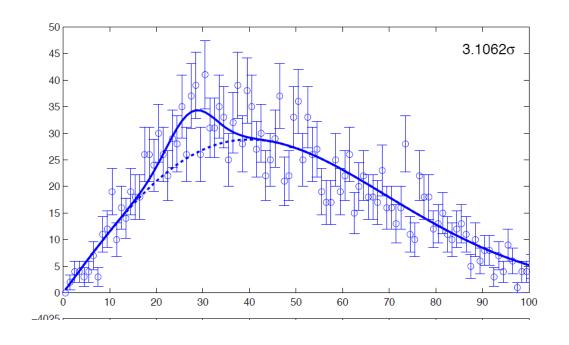
•Or this?



•Or this?

Obviously NOT!

•ALLTHESE
"SIGNALS" ARE BG
FLUCTUATIONS

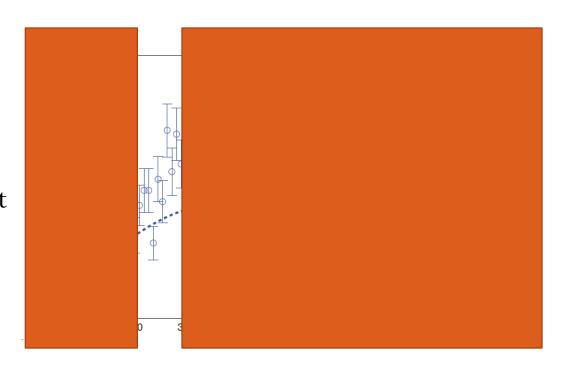


•Having no idea where the signal might be there are two options

•OPTION I:

scan the mass range in pre-defined steps and test any disturbing fluctuations (do not confuse me with the facts)

Perform a fixed mass analysis at each point



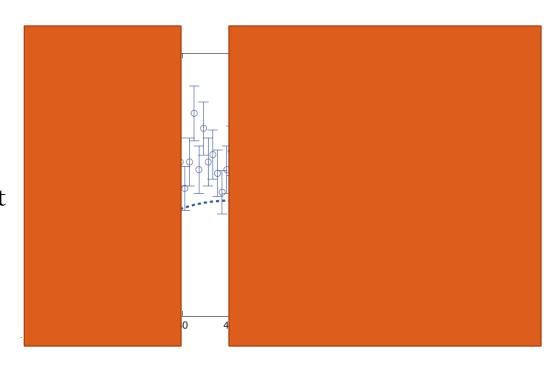
$$q_{fix,obs}(\hat{\mu}) = -2 \ln \frac{L(b)}{L(\hat{\mu}s(m) + b)}$$

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

•OPTION I:

scan the mass range in pre-defined steps and test any disturbing fluctuations (do not confuse me with the facts)

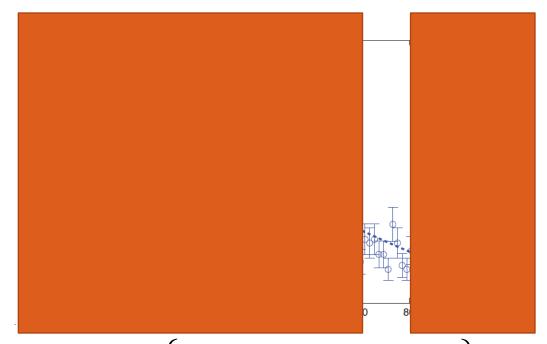
Perform a fixed mass analysis at each point



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

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

•Assuming the signal can be only at one place, pick the one with the smallest p-value (maximum significance)



$$q = \max_{m} \left\{ q_{fix,obs}(\hat{\mu}) \right\} = \max_{m} \left\{ -2 \ln \frac{L(b)}{L(\hat{\mu}s(m) + b)} \right\}$$
$$= \min_{m} \left\{ p - value \right\}$$

Look Elsewhere Effect: Floating Mass

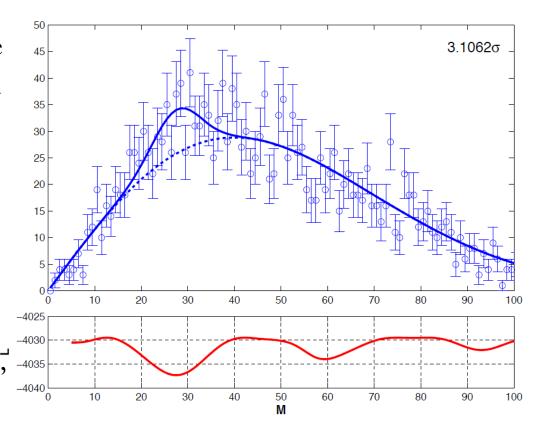
•Option II:

Leave the mass floating

•Having no idea where the signal might be you would allow the signal to be anywhere in the **search** range and use a modified test statistic

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

•For the same observation, -4035 the p-value increases because more possibilities are opened





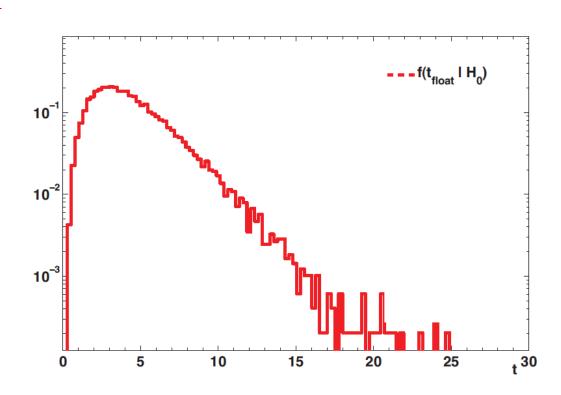
the test statistis

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

•The null hypothesis PDF

$$f(q_{float} | H_0)$$

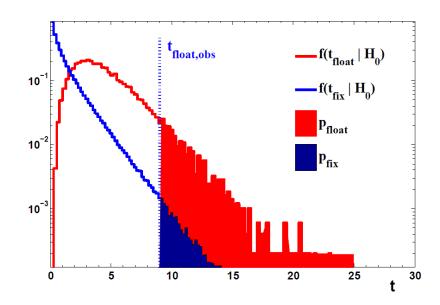
does not follow a chi-squared with 2dof because there are multiple minima depending on the size of the search range and resolution



trial#

- •Assume a maximal local fluctuation at mass $\hat{m} = 30$
- •We can calculate the following p-value

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



$$p_{fix} = \int_{q_{obs}} f(q_{fix} | H_0) dq_{fix} < p_{float} = \int_{q_{obs}} f(q_{float} | H_0) dt_{float}$$

$$trial # = \frac{\int_{q_{obs}} f(q_{float} | H_0) dt_{float}}{\int_{q_{obs}} f(q_{fix} | H_0) dt_{fix}} = \frac{p_{float}}{p_{fix}} > 1$$

Can we analytically calculate the trial# (or p_float)?



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

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

under the null hypothesis

• 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 $\hat{q}_0 \equiv q_0(\hat{\theta}) = -2\log\frac{\mathcal{L}(\mu=0)}{f(\hat{\mu},\hat{\theta})} = \max_{\theta}[q_0(\theta)]$

for which we want to calculate p-value= $P(\max_{\theta}[q_0(\theta)] \ge u), u = Z^2$



The profile-likelihood test statistic

(with a nuisance parameter that is not defined under the Null hypothesis)

Consider the teststatistic:

$$q_0(\theta) = -2\log\frac{\mathcal{L}(\mu=0)}{\mathcal{L}(\hat{\mu},\theta)} \quad \begin{array}{ll} H_0: \mu=0 & \quad \mu=\text{``signal strength''} \\ H_1: \mu>0 \end{array}$$

- For some fixed θ , $q_0(\theta)$ has (asymptotically) a chi² distribution with one degree of freedom by Wilks' theorem.
- $q_0(\theta)$ is a <u>chi² random field</u> over the space of θ (a random variable indexed by a continuous parameter(s)). we are interested in

$$\hat{q}_0 \equiv q_0(\hat{\theta}) = -2\log\frac{\mathcal{L}(\mu=0)}{\mathcal{L}(\hat{\mu},\hat{\theta})} = \max_{\theta}[q_0(\theta)] \qquad \text{is the global maximum point}$$

For which we want to know what is the p-value

p-value=
$$P(\max_{\theta}[q_0(\theta)] \ge u), u = Z^2$$

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The profile-likelihood test statistic

(with a nuisance parameter that is not defined under the Null hypothesis)

Usually we only look for 'positive' signals
 (downward fluctuations of the BG are not considered as
 evidence against the BG)

$$q_{0}(\theta) = \begin{cases} -2\log\frac{\mathcal{L}(\mu = 0)}{\mathcal{L}(\hat{\mu}, \theta)} & \hat{\mu} > 0 \\ 0 & \hat{\mu} \leq 0 \end{cases} \qquad \text{$q_{0}(\theta)$ is `half chi²'} \\ 0 & \hat{\mu} \leq 0 \qquad \text{[H. Chernoff, Ann. Math. Stat. 25, 573578 (1954)]} \end{cases}$$

The p-value just get divided by 1/2

• Or equivalently consider $\hat{\mu}$ as a gaussian field

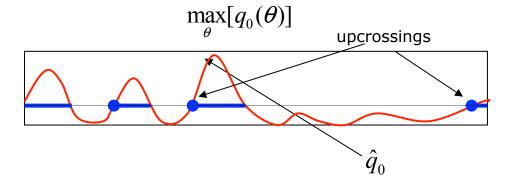
(since
$$q_0(\theta) = \left(\frac{\hat{\mu}(\theta)}{\sigma}\right)^2$$
)



Random fields (1D)

 In 1 dimension: points where the field values become larger then u are called upcrossings.





• The probability that the global maximum is above the level u is called exceedance probability. (p-value of \hat{q}_0)

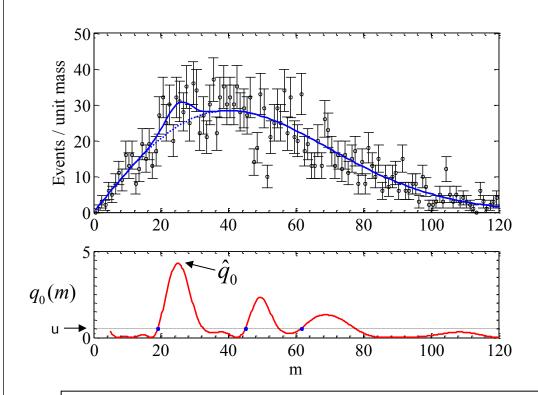
$$P(\max_{\theta}[q_0(\theta)] \ge u)$$

Random fields

- Fortunately, quite a lot of statistical literature on the properties of random fields
- [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.

Applications in Cosmology, Brain mapping, Oceanography ...

The 1-dimensional case



For a chi² random field, the expected number of upcrossings of a level *u* is given by: [Davies,1987]

$$E[N_u] = \mathcal{N}_1 e^{-u/2}$$

Note the inequality:

$$E[N_u] \ge P(N_u > 0)$$

To have the global maximum above a level u:

- Either have at least one upcrossing $(N_u>0)$ or have $q_0>u$ at the origin $(q_0(0)>u)$:



$$P(\hat{q}_0 > u) \le P(N_u > 0) + P(q_0(0) > u)$$

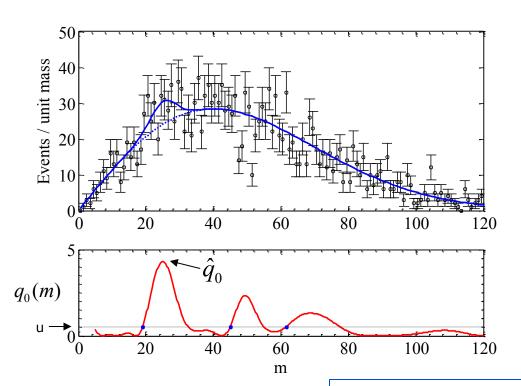
$$\le E[N_u] + P(q_0(0) > u)$$

[R.B. Davies, Hypothesis testing when a nuisance parameter is present only under the alternative. Biometrika **74**, 33–43 (1987)]

Becomes an equality for large *u*



The 1-dimensional case



$$E[N_u] = \mathcal{N}_1 e^{-u/2}$$

The only unknown is \mathcal{N}_1 , which can be estimated from the average number of upcrossings at some low reference level

$$E[N_u] = \mathcal{N}_1 e^{-u/2}$$

$$E[N_{u_0}] = \mathcal{N}_1 e^{-u_0/2} \Longrightarrow$$

$$\mathcal{N}_1 = E[N_{u_0}] e^{u_0/2}$$

The p-value can then be estimated by Davies' formula

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$$P(q_0 > u) \le \mathcal{N}_1 e^{-u/2} + \frac{1}{2} P(\chi_1^2 > u) \Longrightarrow$$

$$P(q_0 > u) \le E[N_{u_0}]e^{(u_0 - u)/2} + \frac{1}{2}P(\chi_1^2 > u)$$

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







Largest significance

 $m_{\star} \sim 750 \text{GeV}, \Gamma_{\star} \sim 45 \text{GeV}(6)$

Local $Z = 3.9\sigma$

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

with m=500-2000 will draw our attention

$$P_{global}(u) \approx p_{local}(u) + E(n_{u_0})e^{\frac{u_0 - u}{2}}$$
 $p_{local} = 5.10^{\circ}$

$$u_0 = 0$$

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

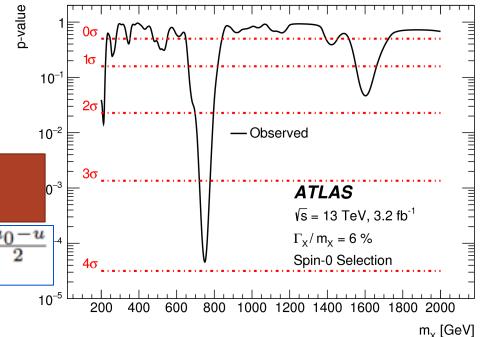
$$u = Z^2 = 3.9^2 = 15.2$$

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

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

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

Eilam Gross, WIS



Trial Factor

• The Trial factor is given by

$$u = Z^{2}$$

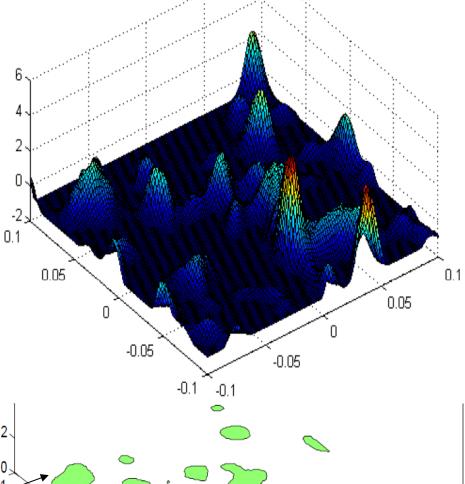
$$trial \# = \frac{p_{global}}{p_{local}} \le \frac{\mathcal{N}_{1}e^{-u/2} + \frac{1}{2}P(\chi_{1}^{2} > u)}{\frac{1}{2}P(\chi_{1}^{2} > u)}$$

$$trial \#(u >> 1) = 1 + \sqrt{\frac{\pi}{2}}\mathcal{N}_{1}Z_{fix} \approx \sqrt{\frac{\pi}{2}}\mathcal{N}_{1}Z_{fix}$$

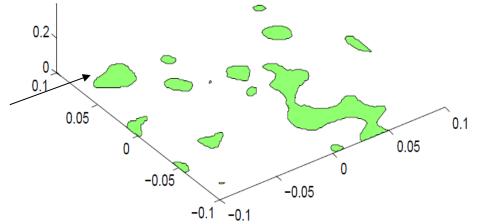
 $\mathcal{N}_{\scriptscriptstyle 1}$ Is the number of independent search regions

Random fields (>1 D)

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

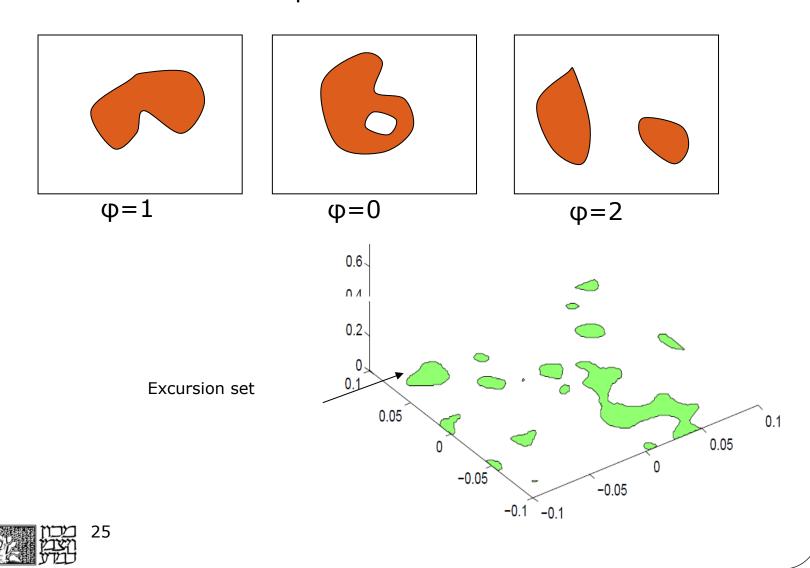


Excursion set



Euler characteristic

Number of disconnected components minus number of `holes'



The *n*-dimensional case

 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

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

Here:

 A_{u} is the excursion set of the field above a level u (set of points where $q_0(\theta) > u$) $\phi(A_u)$ is it's Euler characteristic ρ_d are 'universal' functions (depend only on the level u and the type of distribution)

e.g. for a chi² field with s degrees of freedom:

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

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

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

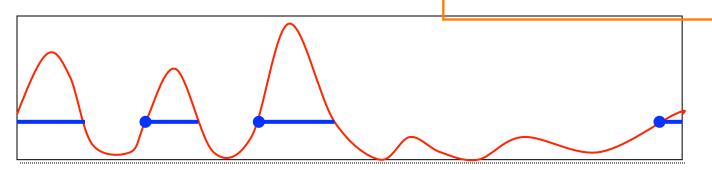
[R.J. Adler and J.E. Taylor, Random Fields and Geometry (2007)

Spring photophological in Mathematics]



Euler characteristic

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



In 1 dimension:

$$\begin{split} \varphi(A_u) &= N_u + \mathbf{1}_{[q_0(0) > u]} \\ E[\varphi(A_u)] &= E[N_u] + \mathrm{P}(q_0(0) > u) \\ &= \mathcal{N}_1 e^{-u/2} + \frac{1}{2} P(\chi_1^2 > u) \\ &\quad \text{(Davies' Bound)} \end{split}$$

The general case

$$\mathcal{N}_0 = ext{Euler characteristic of the } \
ho_0(u) = ext{P}(q_0 > u)$$

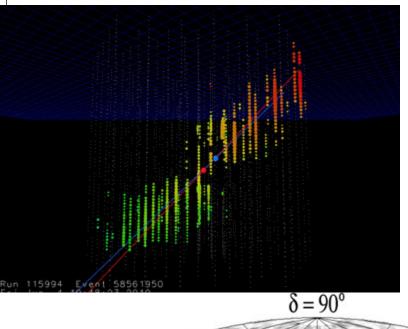
In general for high-level excursions

$$E[\varphi(A_u)] \to P(\max_{\theta}[q_0(\theta)] \ge u)$$



[J. Taylor, A. Takemura, and R. J. Adler, Ann. Probab. 33, 4 (2005)]

2-D example: IceCube search for astrophysical neutrino point sources



IceCube looks for neutrino sources,

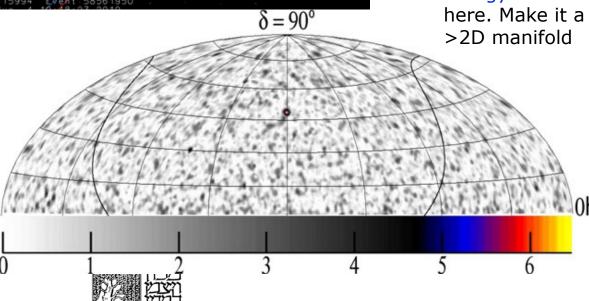
2-D Search over the sky (θ, ϕ)

Unbinned likelihood:

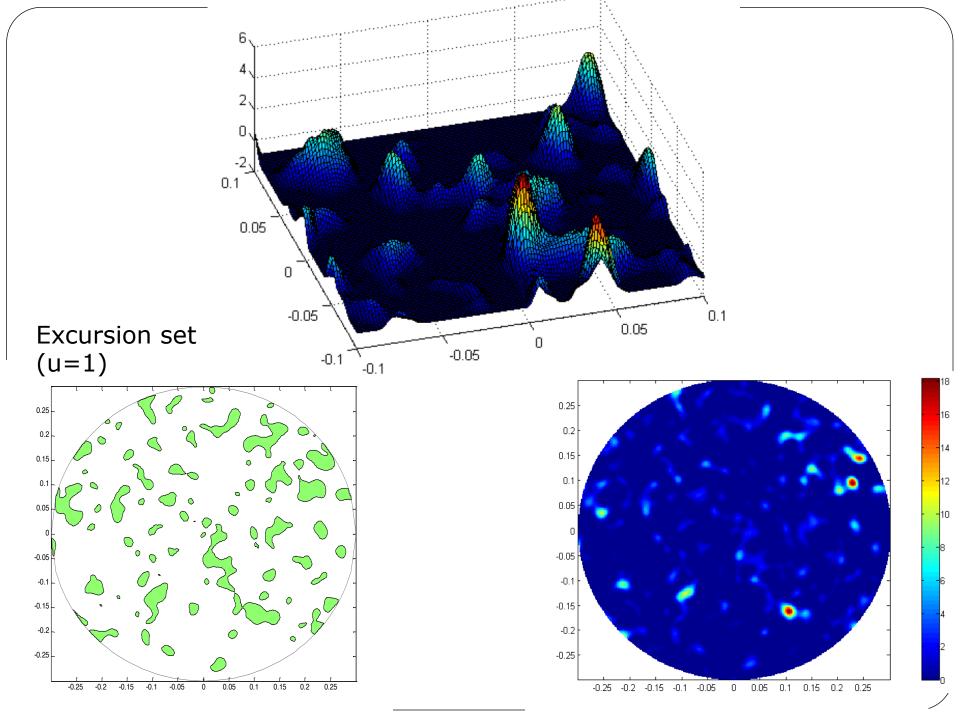
$$\mathcal{L}(\vec{x}_s, n_s) = \prod_i \left(\frac{n_s}{N} f_s(x_i) + (1 - \frac{n_s}{N}) f_b(x_i) \right)$$

$$\vec{x}_{s} = (\theta, \varphi)$$

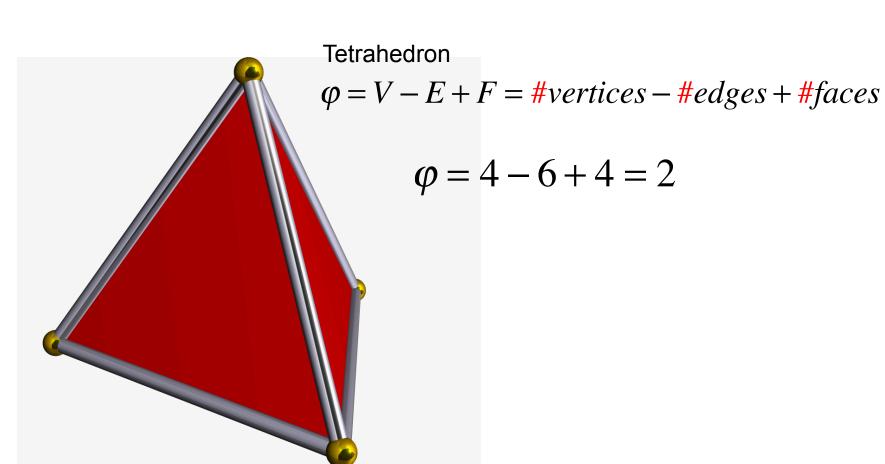
Signal parameters can also include energy and time, not considered here. Make it a multidomensional >2D manifold



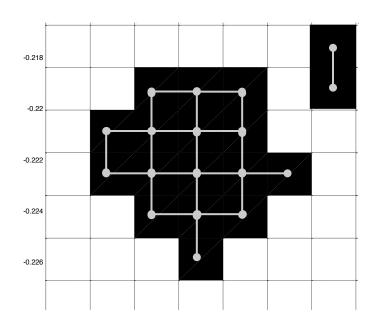
J. Braun, J. Dumm, F. De Palma, C. Finley, A. Karle, and T. Montaruli, Astropart. Phys. 29, 299 (2008); [arXiv:0801.1604]



Calculation of the Euler characteristic



Calculation of the Euler characteristic



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

$$\phi = 18(points) - 23(edges) + 7(faces)$$
= 2

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

For a chi² field in 2 dimensions:

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

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

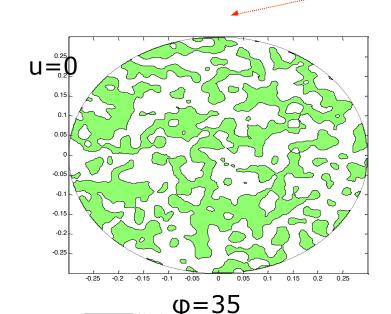
From 20 bkg. Simulations:

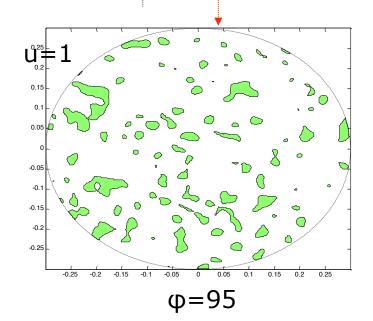
$$\langle \varphi_0 \rangle = 33.5 \pm 2$$

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

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

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



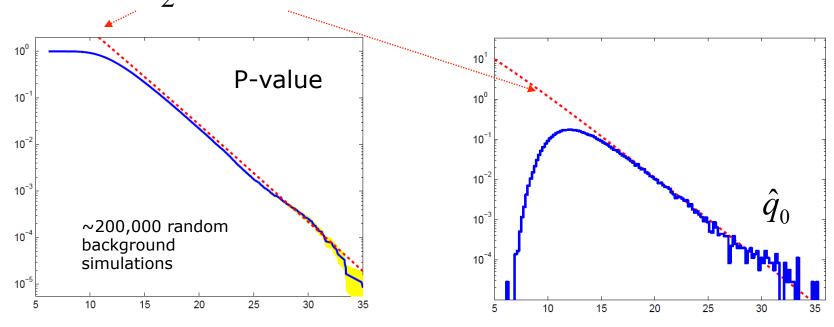


2-d example: search for neutrino sources (IceCube)

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

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

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



e.g.:
$$P(max q_0>30) = (2.5 \pm 0.4)x10^{-4}$$
 (estimated)
E.C. Formula : $(2.28 \pm 0.06)x10^{-4}$



Slicing

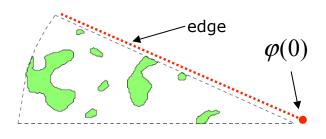
• Exploit the azimuthal angle symmetry to reduce computations: $\phi = 0 = 1 + 1 - 2$

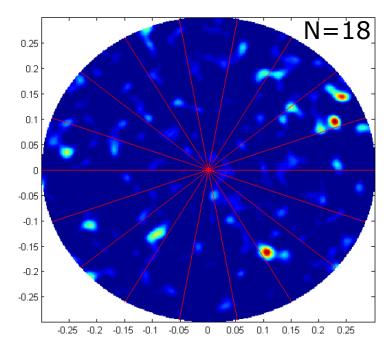
$$\varphi(A \cup B) = \varphi(A) + \varphi(B) - \varphi(A \cap B)$$

Divide to N slices:

$$\varphi = \sum_{i} [\varphi(\text{slice}_{i}) - \varphi(\text{edge}_{i})] + \varphi(0)$$

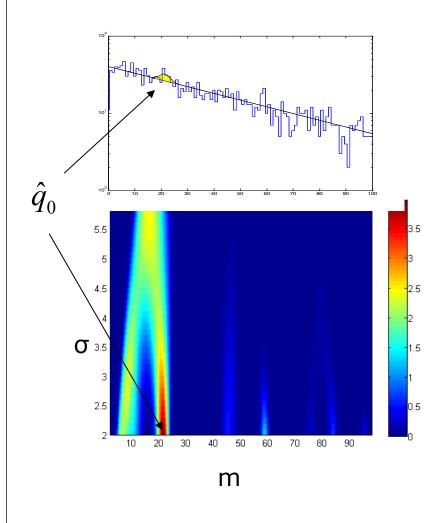
$$E[\varphi] = N \times (E[\varphi(\text{slice})] - E[\varphi(\text{edge})]) + \varphi(0)$$







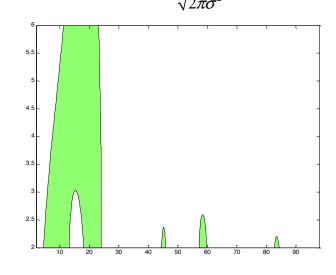
2-D exapmle #2: resonance search with unknown width



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

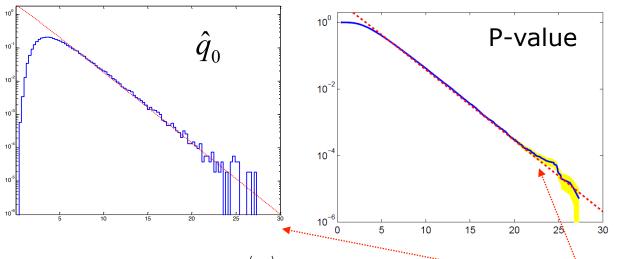
$$\mathcal{L} = \prod_{i} \frac{N_{s} f_{s}(x_{i}) + N_{b} f_{s}(x_{i})}{N_{s} + N_{b}} \times Poiss(N \mid N_{s} + N_{b})$$

$$f_{s}(x; m, \sigma) = \frac{1}{\sqrt{2\pi\sigma^{2}}} e^{\frac{-(x-m)^{2}}{2\sigma^{2}}} \qquad f_{b}(x) = ce^{-cx}$$

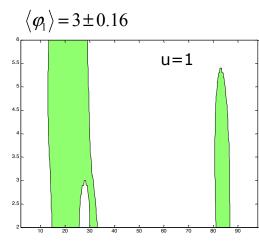


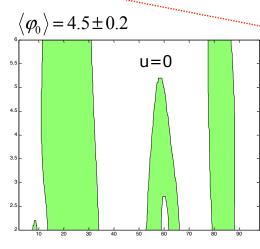


2-D exapmle #2: resonance search with unknown width



Excellent approximation above the ~2σ level



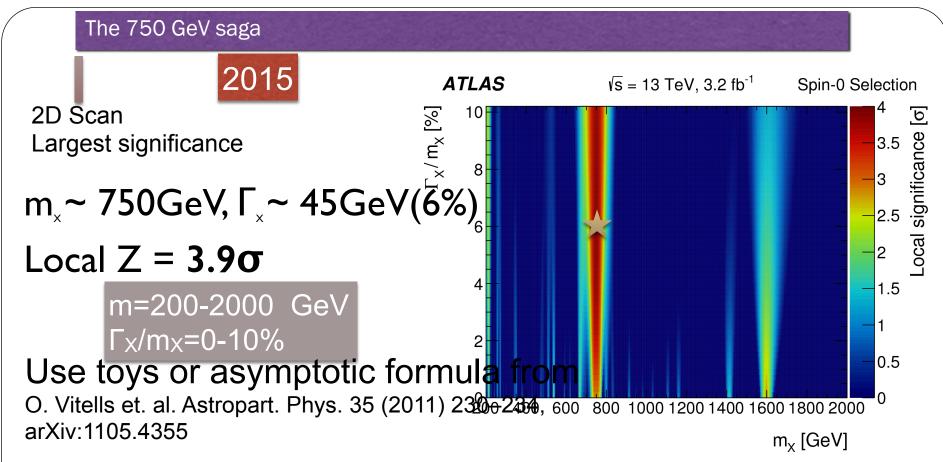


$$E[\varphi(A_u)] = \frac{1}{2} P(\chi^2 > u) + (\mathcal{N}_1 + \mathcal{N}_2 \sqrt{u}) e^{-u/2}$$

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

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





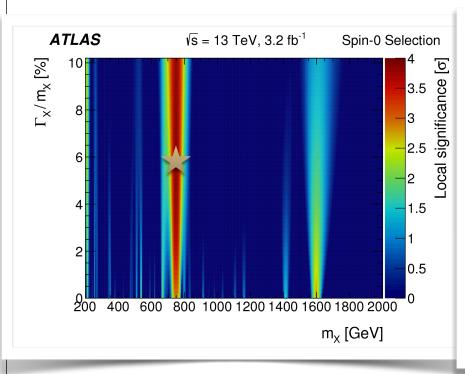
$$Z_{local} = 3.9\sigma$$

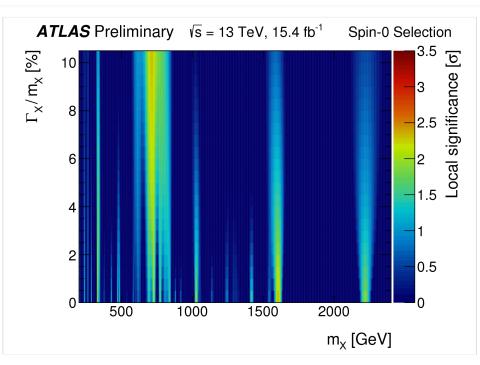
$$Z_{olobal} = 2.1\sigma$$

 $Z_{global} = 2.1\sigma$ 2.1 σ is not something to write home about

2015 vs 2016, which is which?

And indeed,local 3.9σ turns into a global 2.1σ which is not something to write home about





Summary

p-value
$$\approx$$

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

- The Euler characteristic formula provides a practical way of estimating the look-elsewhere effect.
- Applicable in wide range of applications, such as astrophysical searches for neutrino sources or resonance search with unknown width, and in any number of search dimensions.
- The procedure for estimating the p-value is simple and reliable.