



Look-elsewhere effect in neutrino oscillation searches:

Phill Litchfield, Abbey Waldron



Introduction

Multiple hypothesis testing
and testing one hypothesis multiple times:
a unified (re)view

Sara Algeri^{1,2}
Highlights of joint works with:
Prof. David van Dyk¹ and Prof. Jan Conrad²

Testing non-nested models

From a new formulation to a well known problem

Model:
 $(1 - \eta)f(y, \alpha) + \eta g(y, \beta)$ with $0 \leq \eta \leq 1$ (13)

(Note: η is labeled 'Tested on the boundary' and β is labeled 'Not defined under H_0 ')

Test:
 $H_0 : \eta = 0$ versus $H_1 : \eta > 0$
similar argument for $H_0 : \eta = 1$ versus $H_1 : \eta < 1$

Note!

Two sides of the same coin

Two approaches, same inference

- Let $l(\eta|\alpha, \beta_r, y)$ be the log-likelihood of the model of reference.
- The respective score function evaluated at H_0 is $S(\beta_r) = \frac{\partial l(\eta|\alpha, \beta_r, y)}{\partial \eta} \Big|_{\eta=0}$
- Take the normalized version of it, i.e.,

$$S^*(\beta_r) = \frac{S(\beta_r)}{\sqrt{\text{cov}(S(\beta_r), S(\beta_r))}} \quad (16)$$

A sufficient condition on $S^*(\beta_r)$ (Berman's condition)

If the covariance function of $S^*(\beta_r)$ satisfies

$$\sup_{|\beta_r - \beta_t| > \tau} |\text{cov}(S^*(\beta_r), S^*(\beta_t))| \log(\tau) \rightarrow 0 \text{ as } \tau \rightarrow +\infty, \quad (17)$$

THEN TESTING ONE HYPOTHESIS MULTIPLE TIMES AND MULTIPLE HYPOTHESES TESTING ARE (APPROXIMATELY) EQUIVALENT!

This result will be discussed in:
S. Algeri, D.A. van Dyk and J. Conrad. *Testing one hypothesis multiple times*. In preparation, 2016.
(Hopefully, available on ArXiv by the end of the summer.)

Phystat-ν [Kashiwa] talk by S. Algeri

- How to frame discrete models (e.g. mass ordering) as a Look-Elsewhere Effect (LEE).

Followed up 2017

- Looks like we could use this approach for T2K MO [ask me over coffee break]
- Wanted to first understand this approach to LEE
- But no time to actually work on it

2018 out of the blue, Abbey contacted to ask if I had any interesting neutrino / computing projects to work on

- Why, yes!** Yes I do.



LEE in discrete tests

If I observe a 3σ deviation from my Null Hypothesis, this means:

If **the Null Hypothesis is true**
Then **the probability of this occurring by chance is 0.27%**

But if I looked at 100 different data sets, how surprised should I be?

If the probability of at least one occurrence in τ trials is P_τ then:

$$P_{100} = 1 - \overline{P_{100}} = 1 - (\overline{P_1})^{100} = 1 - (1 - P_1)^{100} \\ \simeq 100 \times P_1$$

For this example:

a result with 3σ *local* significance
becomes 1.2σ *global* significance

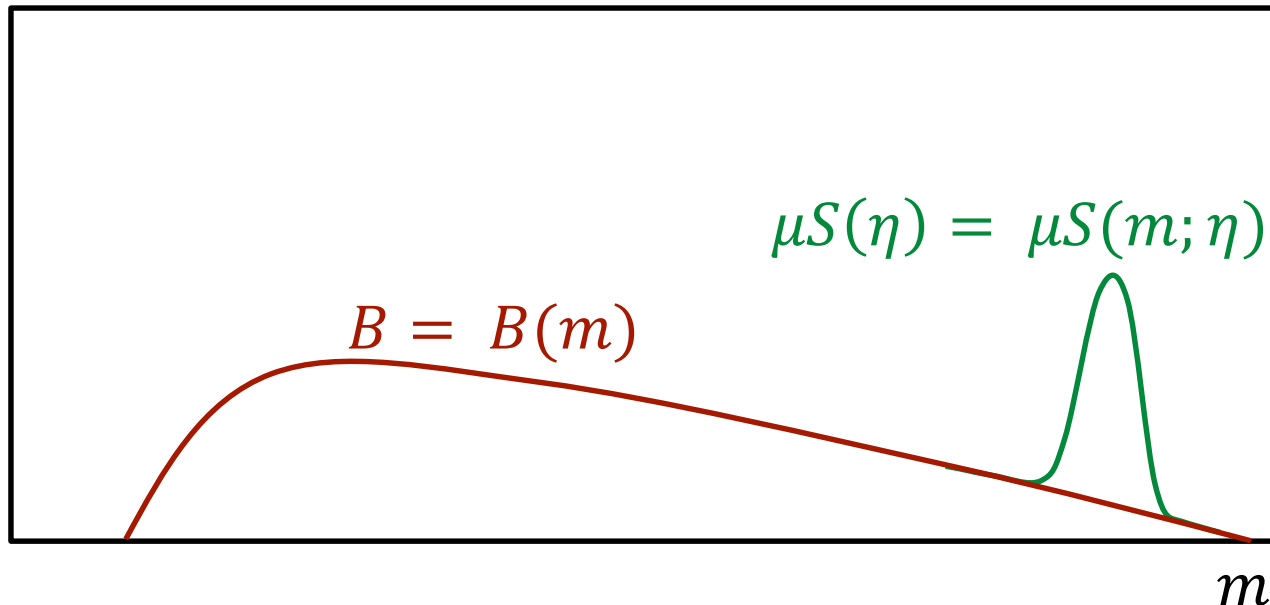
Z_1	Z_{100}
2σ	0.01σ
3σ	1.2σ
4σ	2.7σ
5σ	4.0σ



LEE in continuous tests

Imagine a collider experiment looking for a resonance.

- There is a known **background**
- There might be a **signal** resonance *somewhere* in the search range



⇒ There is a Look-Elsewhere Effect here as well.

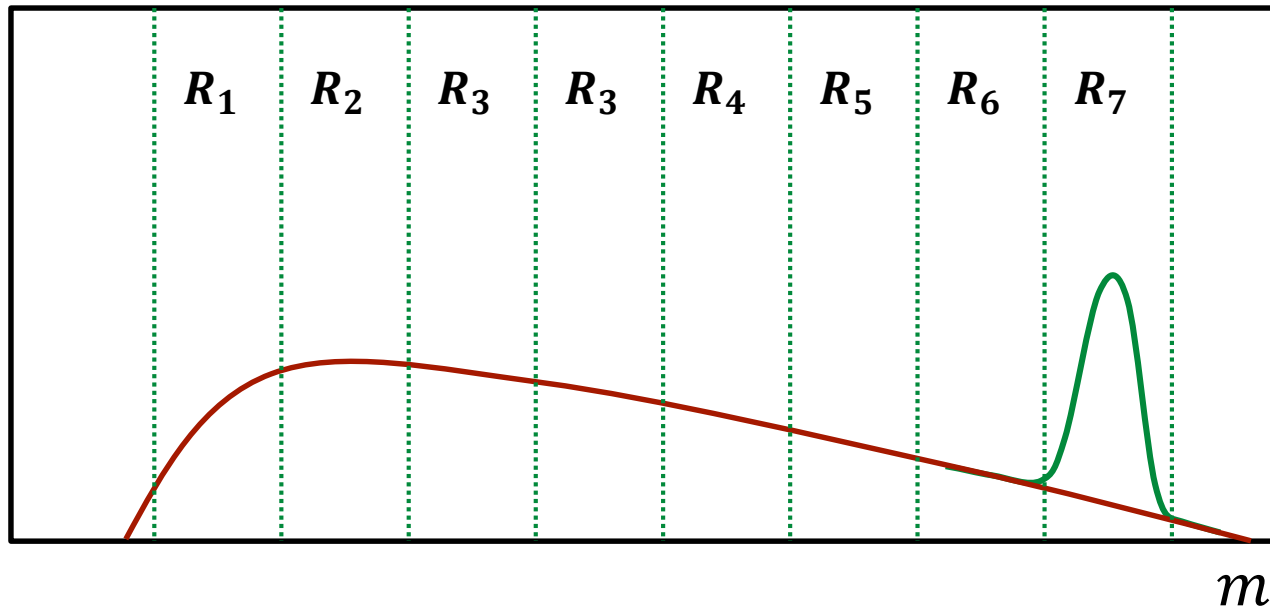
But how to quantify it?



The “Thumb rule”

Quasi-discrete approach: break the search range up into sub-ranges

- Wide enough that ~ 1 resonance can exist within each one...
- Then as in the discrete case $P_\tau = \tau \times P_1$



This **assumes** we are searching at only 7 specific values of η

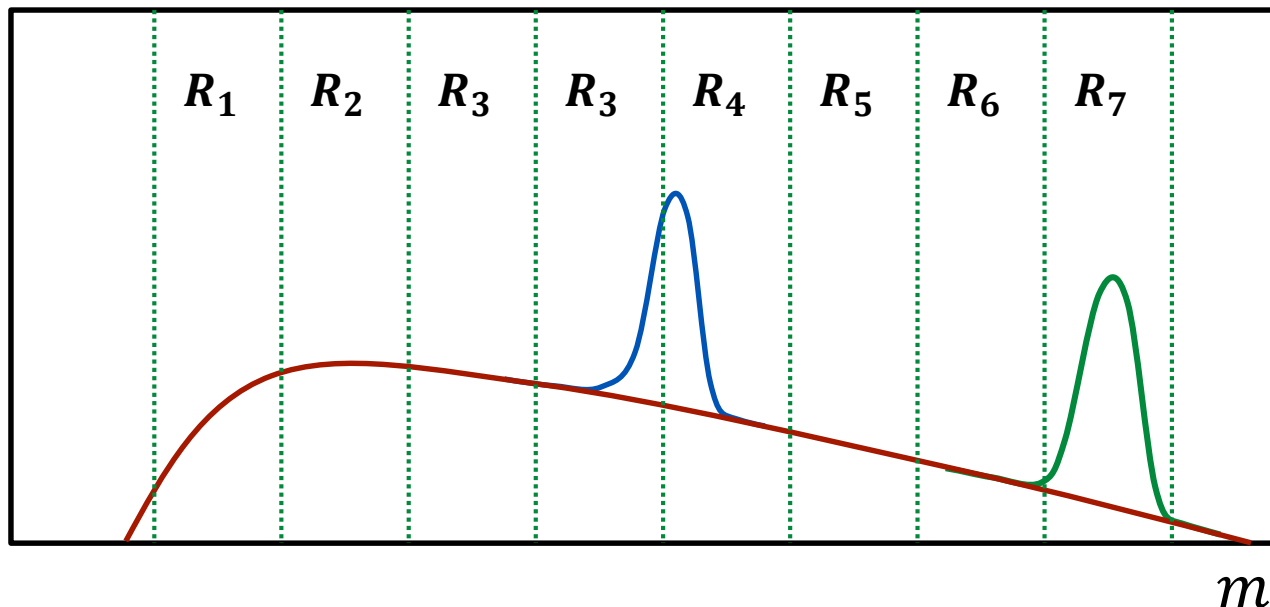
- Still better than ignoring the issue...



The “Thumb rule”

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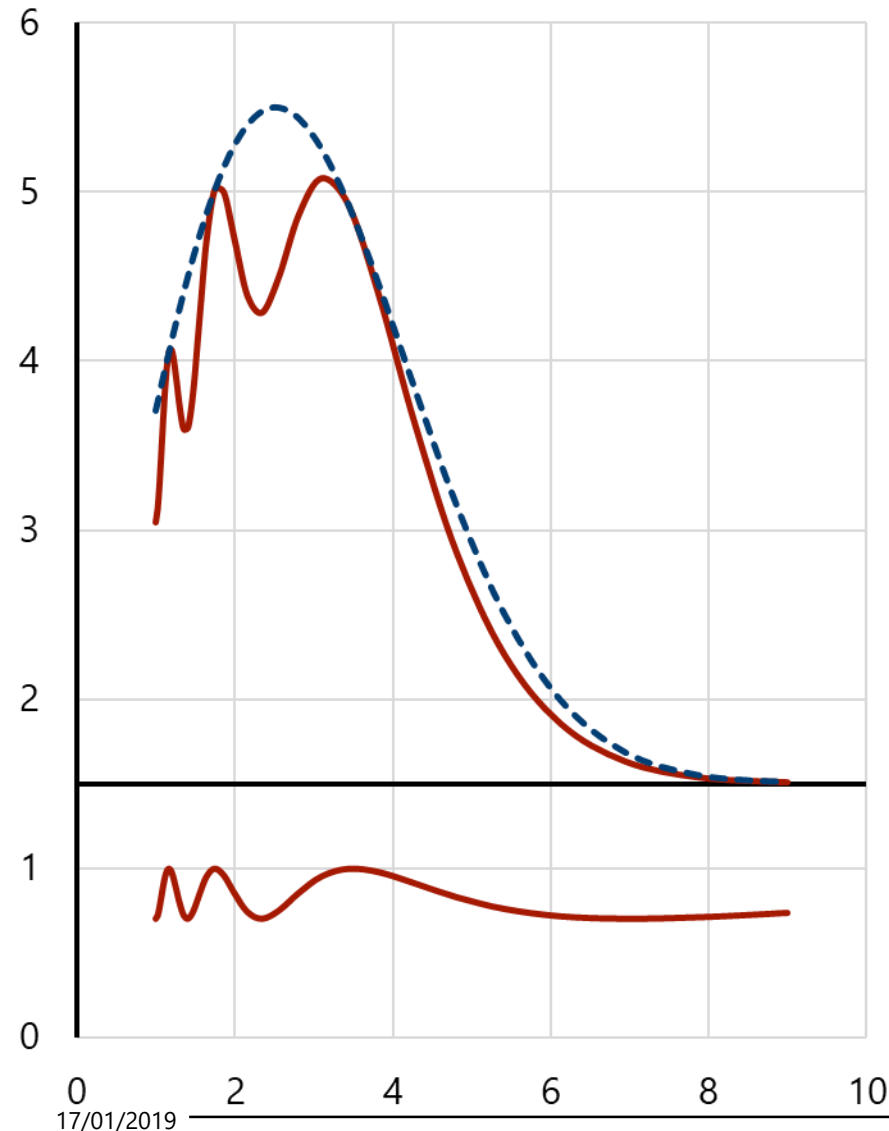


This **assumes** we are searching at only 7 specific values of η

- Still better than ignoring the issue... **but not very realistic**



Neutrino oscillation searches



In searches for new oscillation scales we have:

$$P(\nu_\alpha \rightarrow \nu_\beta) = \delta_{\alpha\beta} \pm \sin^2 2\theta \sin^2 \frac{\Delta m^2 L}{4E}$$

Similar to bump searches:

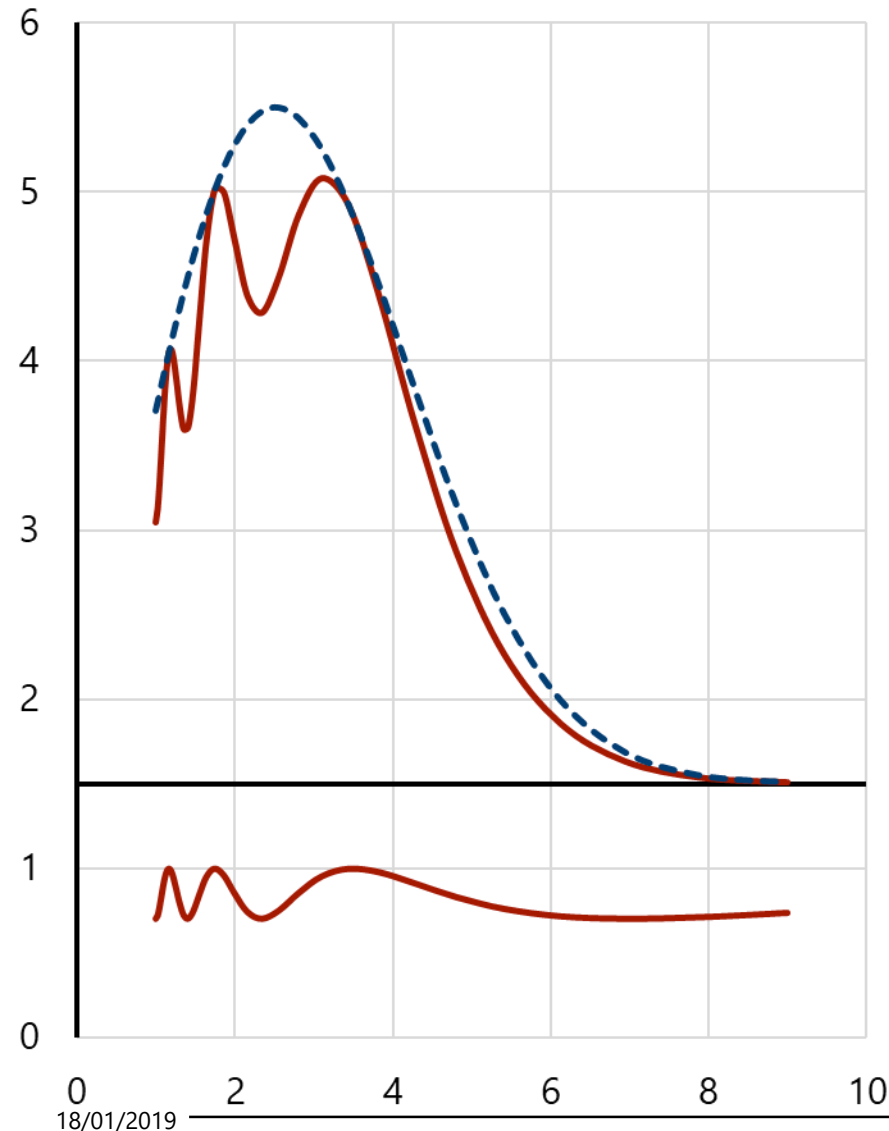
- Location parameter (Δm^2)
- Magnitude ($\sin^2 2\theta$)

But there is a difference:

- The signal is not localised.
- So, how many searches?



How to handle this?



In searches for new oscillation scales we have:

In the neutrino case:

$$P(\nu_\alpha \rightarrow \nu_\beta) = \delta_{\alpha\beta} \pm \sin^2 2\theta \sin^2 \frac{\Delta m^2 L}{4E}$$

- There's no approach equivalent to dividing the spectrum into sub-ranges.

Similar to bump searches:

- Location parameter (Δm^2)
- Magnitude ($\sin^2 2\theta$)
- But the concept of a tuneable search parameter still exists —

**But there is a difference:
how can we use that?**

- The signal is not localised.
- So, how many searches?



Resonance search by scanning the location parameter

Looking for a 'bump' on top of a (known) **background B**

- The bump is a localised feature, parameterised by its **location (η)**, and **magnitude (μ)**

If we already knew to search at η_0 : standard results (Wilks, Chernoff) for significance, based on log-**likelihood ratio $q(\hat{\mu}, \eta_0)$**

But if the search location $\hat{\eta}$ is determined by fitting data, these results will overestimate the significance:

- By definition $q(\hat{\mu}, \hat{\eta}) = \max_{\eta} \{q(\hat{\mu}(\eta), \eta)\} \geq \underbrace{q(\hat{\mu}(\eta_0), \eta_0)}_{\rightarrow \chi^2 \text{ distribution}}$

Physicists: ***"Look-Elsewhere effect"***

Statisticians: ***" η is only present under the alternative hypothesis"***

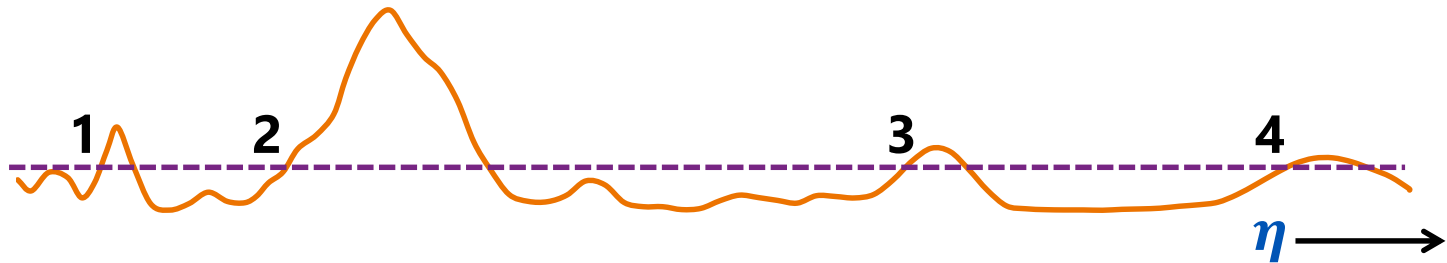
Because of the LEE:

$$P(q(\hat{\eta}) > c) > P(q_{\text{fix}} > c)$$

Davies [1977 & 1987] showed that:

$$P(q(\hat{\eta}) > c) < P(q_{\text{fix}} > c) + \langle N(c) \rangle$$

Where $\langle N(c) \rangle$ is the *expected* number of times q goes above the level c when scanned across η



This is a useful result! Although it is still not exact, it bounds the significance from the other (conservative) side



Gross & Vitells extension

For a χ^2 test: $P(q_{\text{fix}} > c) = P(\chi_s^2 > c)$ and:

Davies 1987

$$\langle N(c) \rangle = \left[\frac{c^{(s-1)} e^{-c}}{\pi 2^s} \right]^{1/2} \frac{1}{\Gamma((s+1)/2)} \int_L^U I(\eta) d\eta$$

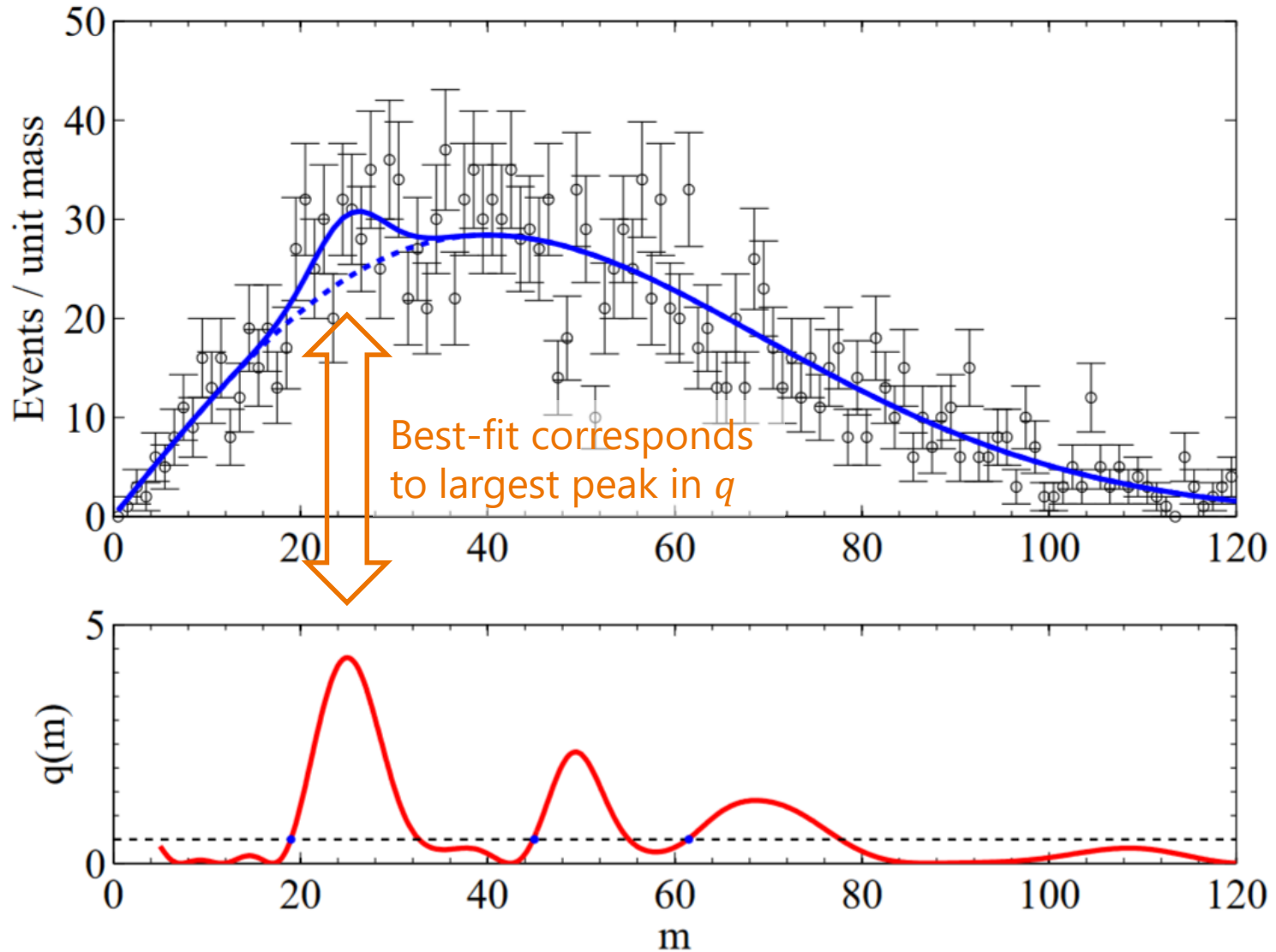
Gross & Vitells [2010] point out that

1. The hard part (the **integral**) is independent of the threshold.
2. The expectation $\langle N(c) \rangle$ can be calculated numerically at some low threshold value (c_0) and evolved to the level of interest (c)

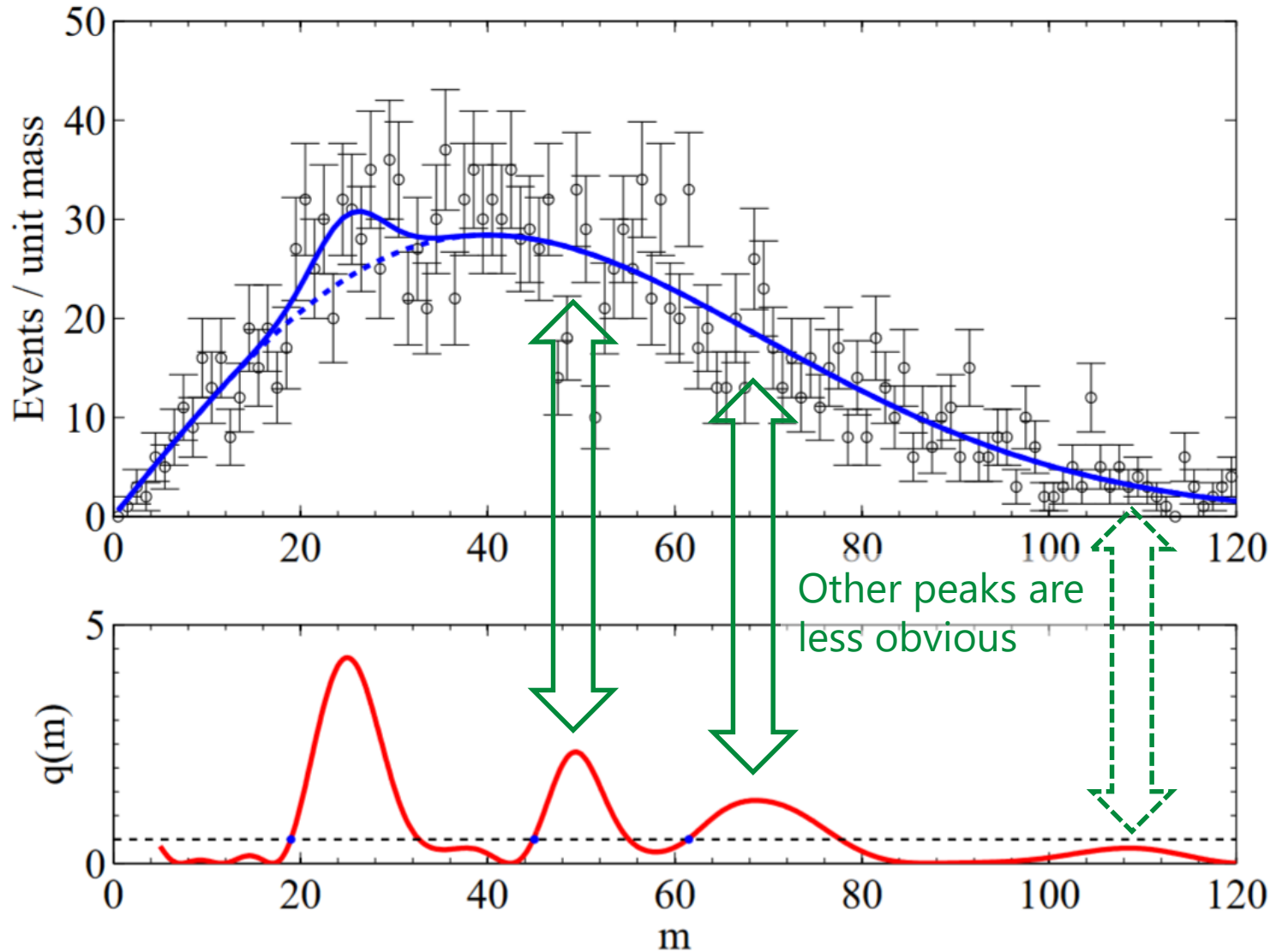
Thus:

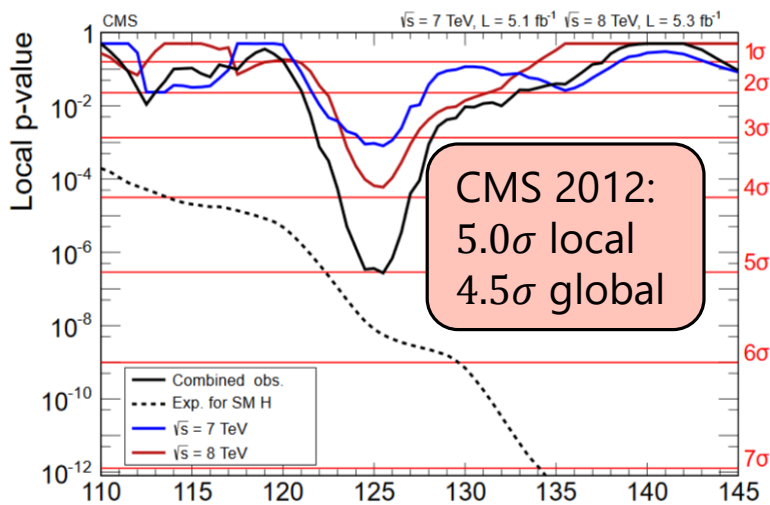
$$P(q(\hat{\eta}) > c) < P(\chi_s^2 > c) + \underbrace{\langle N(c_0) \rangle}_{\text{Evaluate with MC}} e^{-(c-c_0)/2} \underbrace{\left(\frac{c}{c_0} \right)^{(s-1)/2}}_{\substack{s=1 \text{ here} \\ \rightarrow \text{ignore}}}$$

Gross-Vitells example



Gross-Vitells example





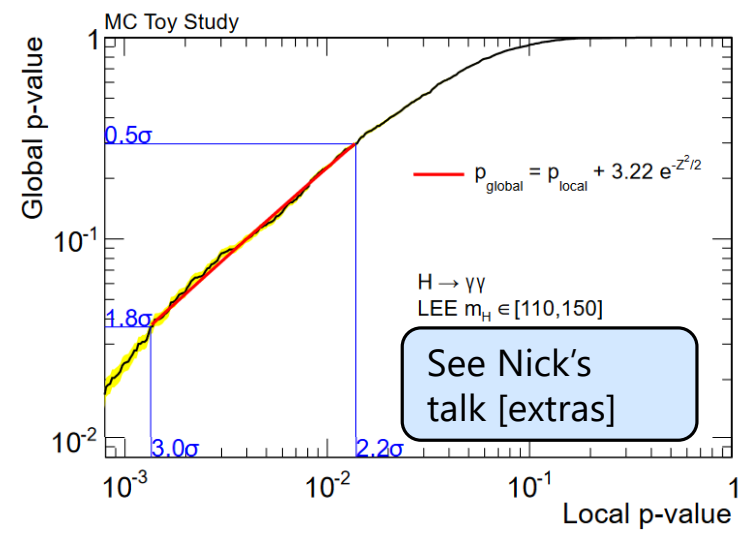
Most LHC resonance searches now use this approach. Written in physicists jargon:

$$P(q(\hat{\eta}) > c) \rightarrow p_{\text{global}}$$

$$P(\chi_s^2 > c) \rightarrow p_{\text{local}}$$

$$c = z^2 \text{ (where } z \text{ is the significance)}$$

$$s = N_{\text{dof}} = 1$$



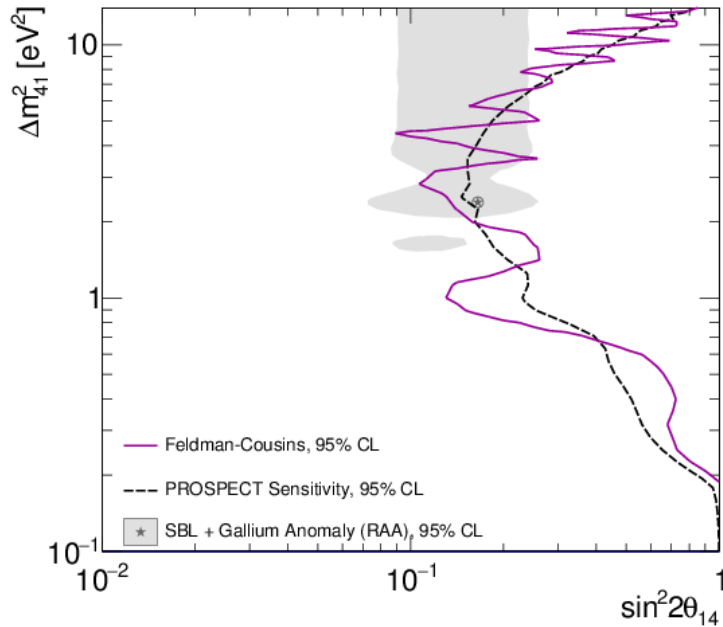
We then have:

$$p_{\text{global}} = p_{\text{local}} + k e^{-z^2/2}$$

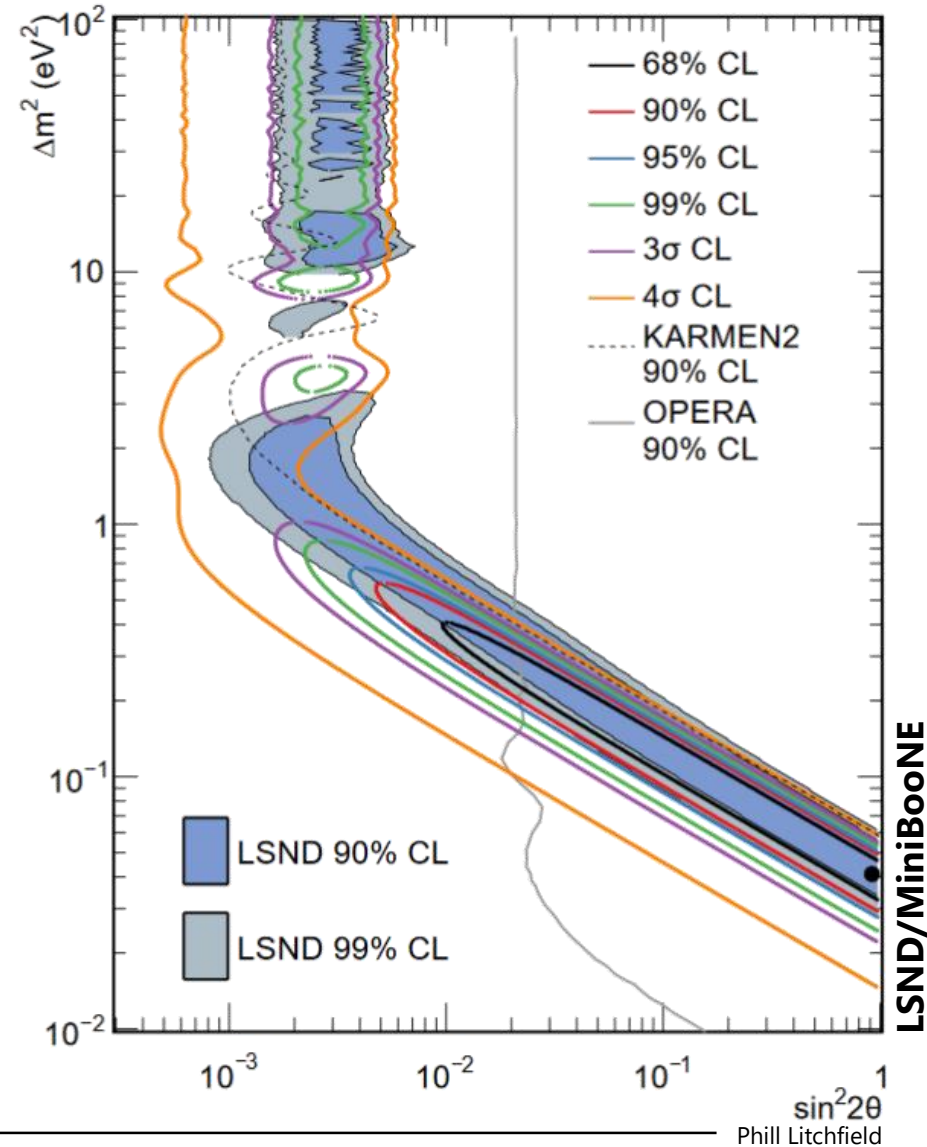
and this z dependency is used to extrapolate from lower-significance MC toys

Sterile neutrino searches

Looking at this approach for neutrino oscillation searches.



Disappearance (◀) or appearance (▶) channels



High energy regime

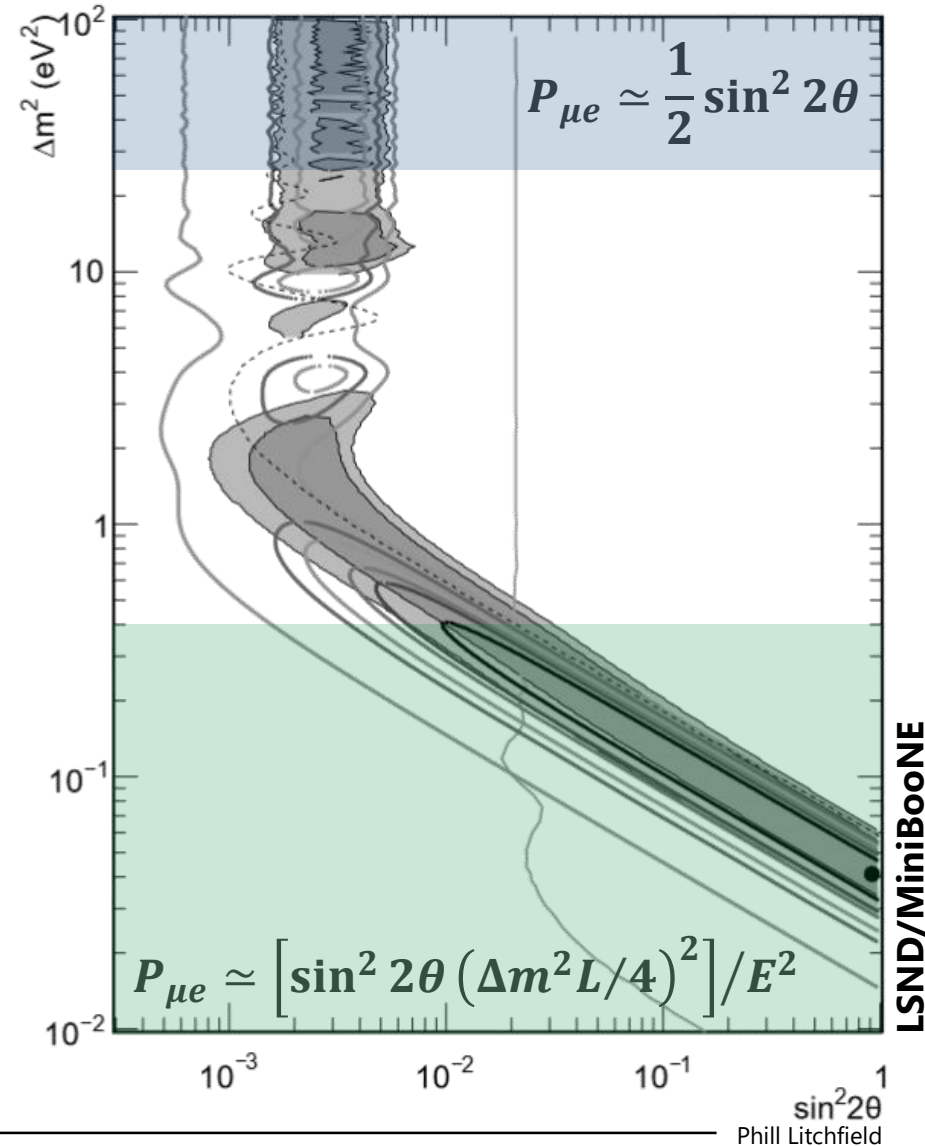
- Finite energy resolution washes out location parameter.

Low energy regime

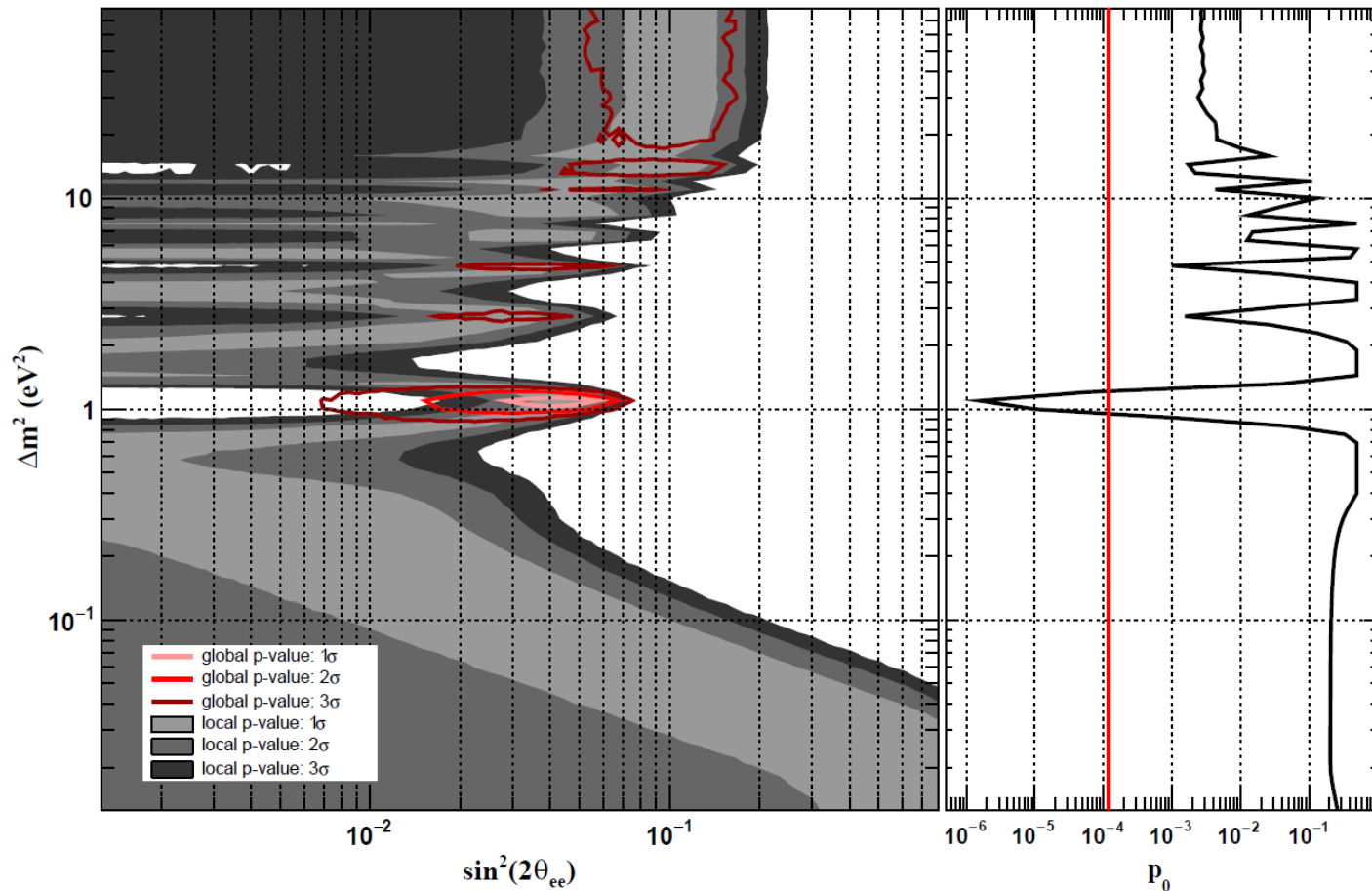
- Location parameter becomes degenerate with the strength parameter

In both cases we transition back to a single degree-of-freedom;

- How will this affect the LEE correction?

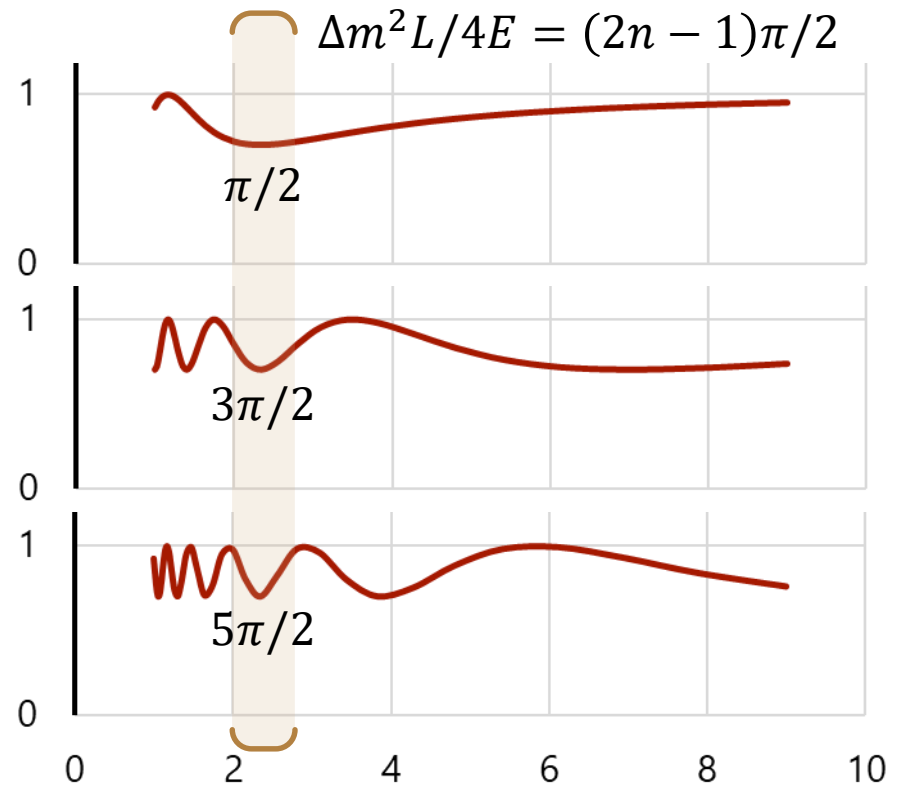
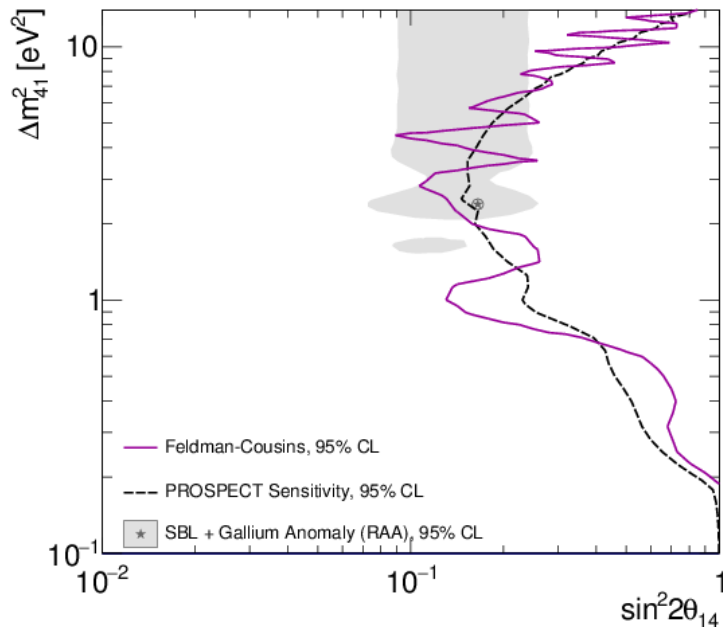


Poster “Statistical methods and issues in sterile neutrino searches” by B.Neumair → Covers similar ideas & connection back to F&C



Behaviour in **high-energy regime** also depends on how "normalisation" is handled.

- Dis/appearance not the (direct) cause



In the **intermediate energy regime**, possible correlation between harmonics?



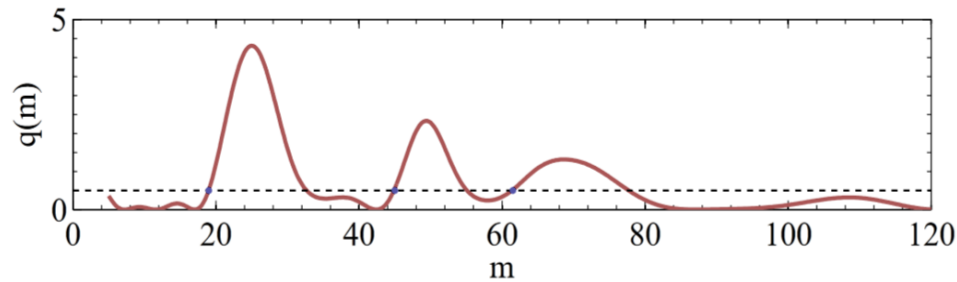
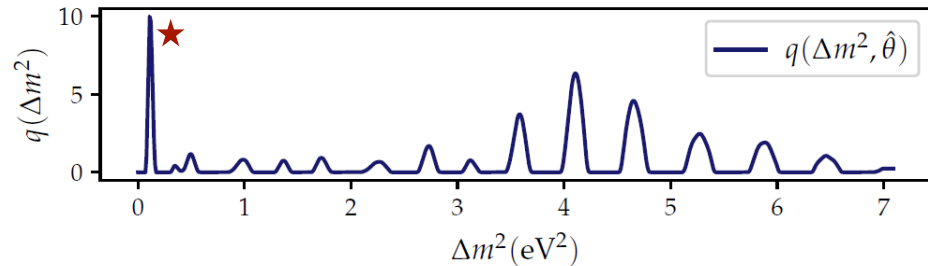
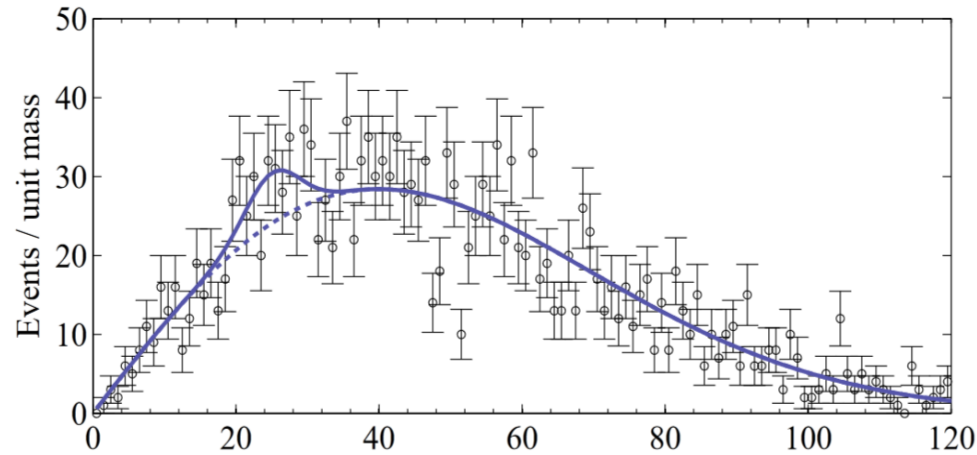
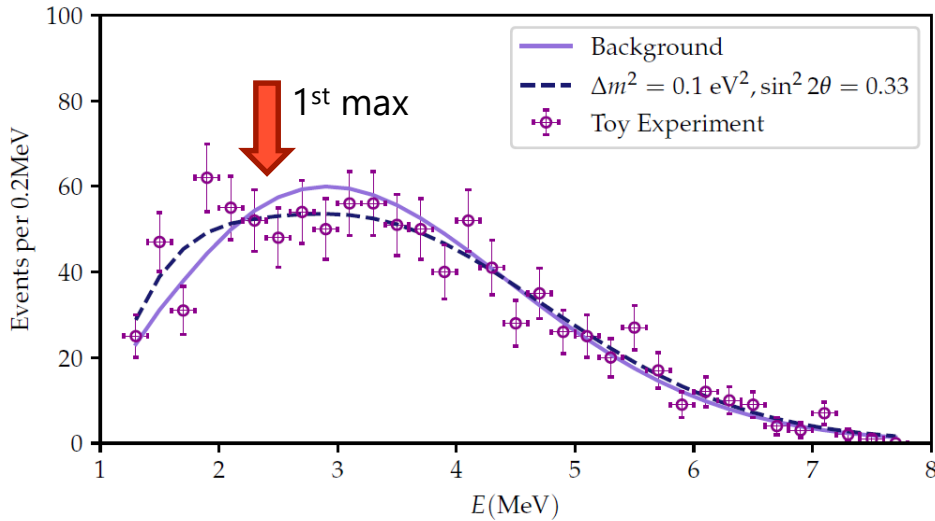
Application to neutrinos

Toy study:

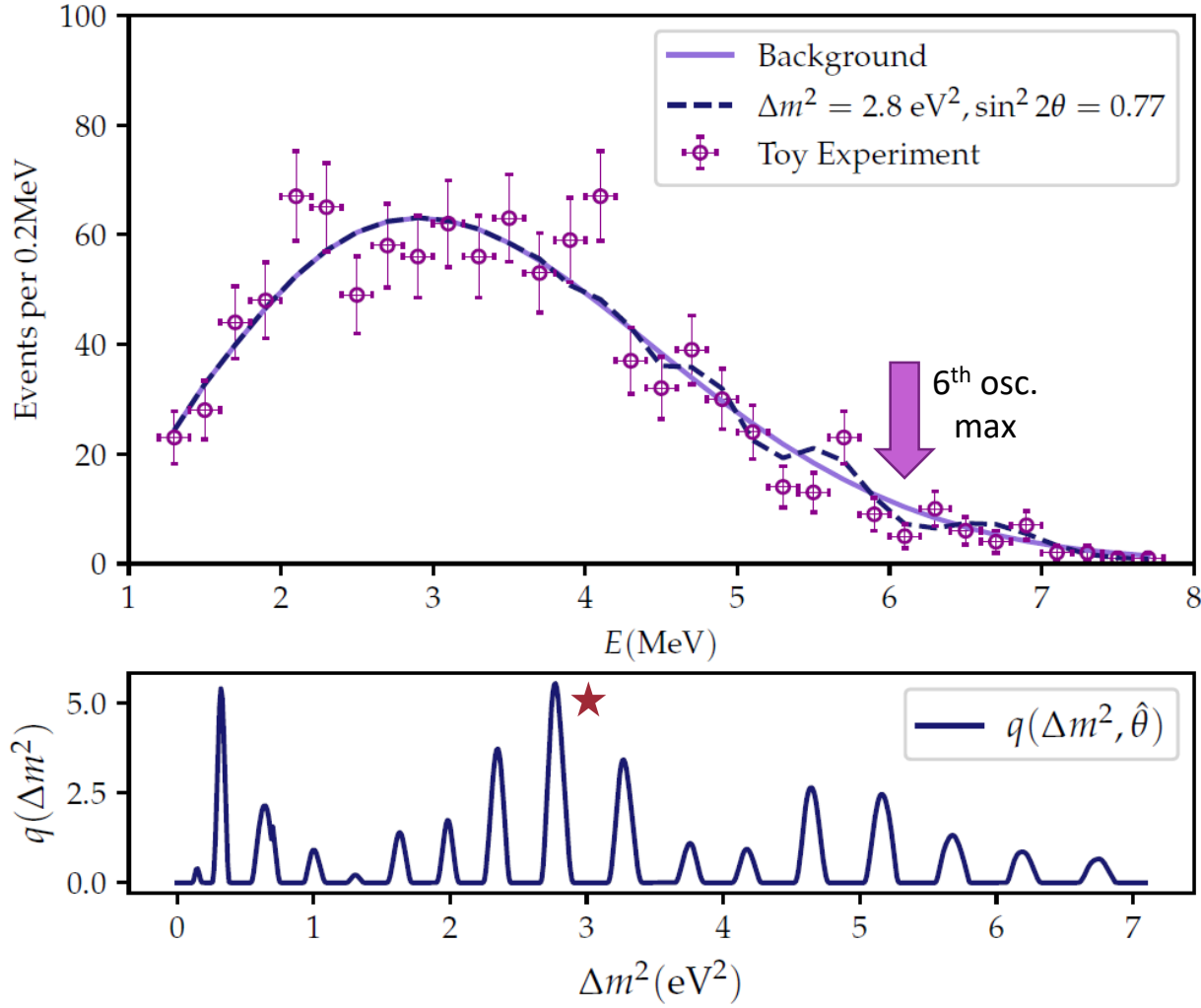
- Point like reactor and detector with 30m baseline
 - [Reality: both reactors & detectors extended over ~m]
- Flux based on RENO 2018 data release
- Fixed 0.2 MeV energy resolution.
 - [Typical: ~10% resolution; so worse >4MeV]
- Free normalisation (i.e. shape only-analysis)
 - Fixed or constrained normalisations have different large Δm^2 behaviour
 - Multi-baseline experiments effectively do shape-only analyses, but several (L,E) schemes in use: ratios, averaging over baselines, ...

Compared to G-V toy study (right), notice:

- No visual mapping to Δm^2
 - Much higher dynamic range
- \swarrow Better resolution
 \searrow Sensitive to 'harmonic' solutions



More examples



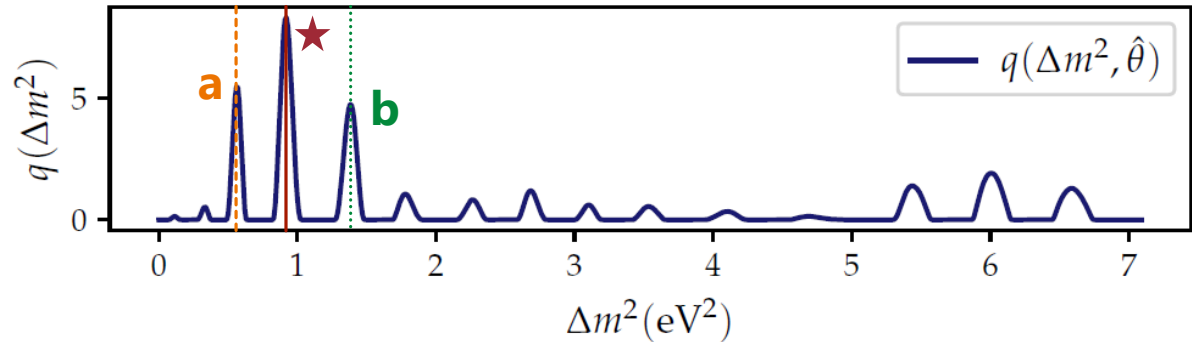
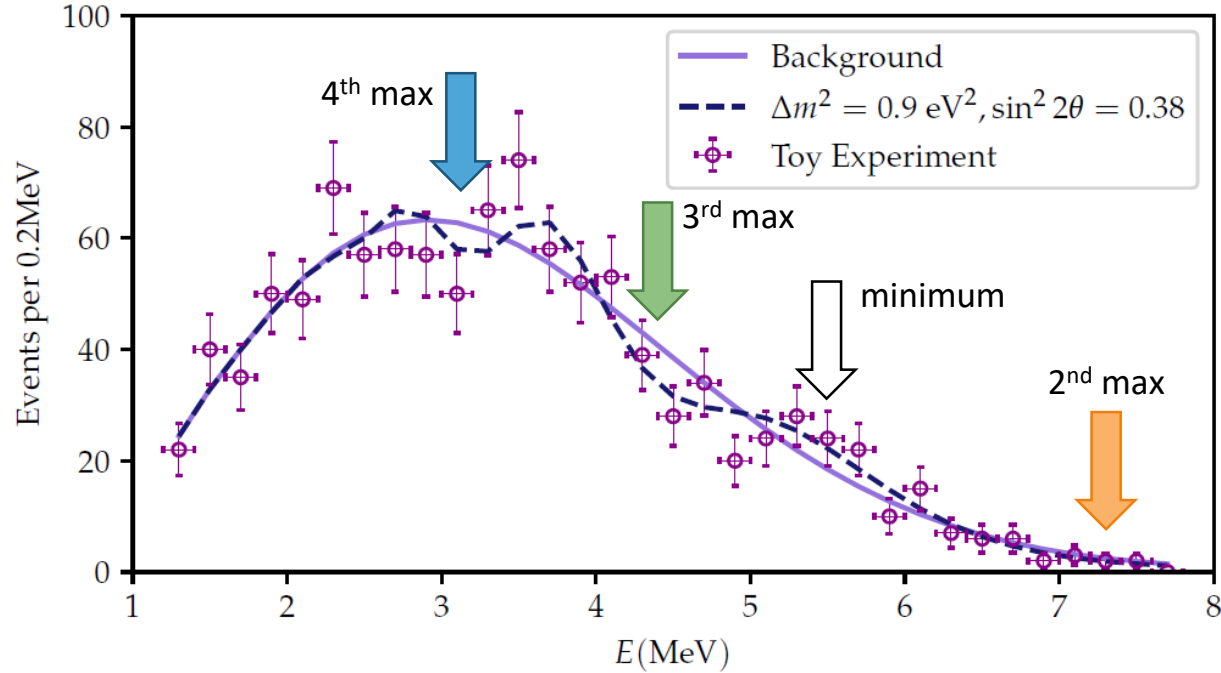
High Δm^2 fits have tail structure – improve E resolution model.

Adjacent harmonic aliases will be at $(n/n + 2)\Delta m^2$

Dips \Rightarrow odd- n
Peaks \Rightarrow even- n

★/**a** $\approx 5/3$ fits the 4.4 MeV dip.

★/**b** $\approx 2/3$ would map a peak to a dip, so this is not an alias.



Sterile neutrino searches should account for the Look-Elsewhere Effect

- Below $\sim 2.5\sigma$, can just do B/G-only toys.
- But not so easy for 'interesting' results
- Note: Better resolution means local value is *more* wrong

Davies / Gross & Vitells method looks usable for estimating global significance for reactor searches

- But not trivial to investigate
- Normalisation, harmonic and small- Δm^2 degeneracies need to be checked & understood, but seem okay so far.

Should be equally viable for (e.g.) SBN appearance at FNAL



Extensions & references

Davis '87 result covers χ^2 -like statistics. If instead use *amplitude* of oscillation as a test statistic, can use similar relationship from Davis '77

Tests of $3+n$ models simply extend this to χ^2 with n d.o.f. Generalised form already exists.

Also want to get back to applying this to MO...

Gross & Vitells: <https://inspirehep.net/record/854732>

Davies '77: Biometrika, 64, 247-254

Davies '87: <https://inspirehep.net/record/854290>

RENO flux: <https://inspirehep.net/record/1676077>

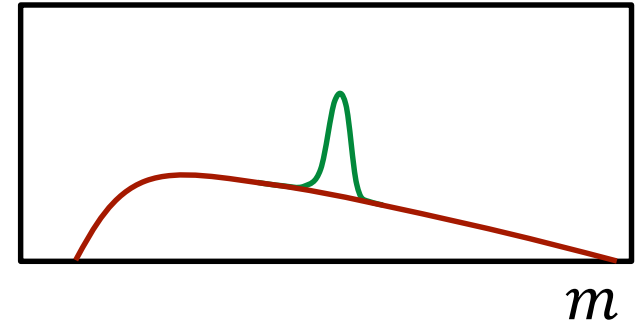


Extra slides

Could test only at wide-spaced values of η , but this is very limiting.

Instead, consider a test statistic $T(\mu, \eta)$.

Most commonly:
$$T = -2 \ln \frac{\mathcal{L}(B)}{\mathcal{L}(\mu S(\eta) + B)}$$



For an *assumed* value η_0

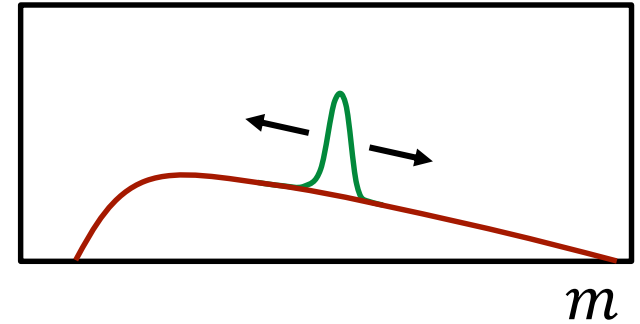
we can find the best fit value $\hat{\mu}(\eta_0)$ that maximises T :

- ✓ Obeys Wilks' theorem \rightarrow significance can be estimated from χ^2
- ✗ Significance only meaningful if the assumption $\eta = \eta_0$ is true.

Could test only at wide-spaced values of η , but this is very limiting.

Instead, consider a test statistic $T(\mu, \eta)$.

Most commonly:
$$T = -2 \ln \frac{\mathcal{L}(B)}{\mathcal{L}(\mu S(\eta) + B)}$$



Instead of assuming η_0

we can find the 2D best fit value $(\hat{\mu}, \hat{\eta})$. But by definition:

$$T(\hat{\mu}, \hat{\eta}) = \max_{\eta} \{T(\hat{\mu}(\eta), \eta)\} \geq T(\hat{\mu}(\eta_0), \eta_0)$$

The RHS obeys Wilks' theorem, so the LHS must not!

In the discrete case we had that $P_\tau \simeq \tau P_1$, (provided $P_1 \ll 1$)

The factor τ is generalised to a non-integer **trial factor**:

$$\tau = \frac{P(T(\hat{\eta}) > c)}{P(T(\eta_0) > c)} = \frac{P(T(\hat{\eta}) > c)}{P(\chi_s^2 > c)} \xrightarrow{s=1} \frac{P(T(\hat{\eta}) > c)}{\sqrt{c}}$$

With which one can convert a “local” significance to a “global” one.

- Or if you prefer, an **effective number of search regions**

The expression derived for this is:

$$\tau_{s=1} \simeq 1 + \frac{\sqrt{c}}{2} \int_L^U I(\eta) d\eta$$

[The integral is deducible from the small threshold $\langle N(c_0) \rangle$]

Testing non-nested models

Blennow M. et al., JHEP, 2014 - Assumptions

...BUT

To make it work we need:

- The x_1, \dots, x_B are independent and approximately Gaussian. (If not independent, the variance is no longer as simple as $4T_0$.)
- We need assumptions similar to those required by Wilks' theorem.

PLUS one (or more) of the following:

- 1 There are no free parameters α and β .
- 2 OR The hyperplanes of the two hypotheses at their minima are parallel, i.e.,

This condition is broken?

$$\left. \frac{\partial f(y_i, \alpha)}{\partial \alpha} \right|_{\alpha=\alpha_{\min}} = \left. \frac{\partial g(y_i, \beta)}{\partial \beta} \right|_{\beta_{\min}}$$

This is very restrictive!

E.g.: Let $f(\alpha) = \alpha$, $\alpha \in [0; 10]$, $g(\beta) = \beta^2$, $\beta \in [15; 20]$ and $\beta_{\min} = 1$

$$\left. \frac{dN}{d\delta_{CP}} \right|_{NO} \neq \left. \frac{dN}{d\delta_{CP}} \right|_{IO}$$

$$1 = \left. \frac{\partial f(\alpha)}{\partial \alpha} \right|_{\forall \alpha \in [0;10]} \neq \left. \frac{\partial g(\beta)}{\partial \beta} \right|_{\beta=1} = 2 \Rightarrow \text{NOT OK.}$$

- 3 OR if T_0 much greater than the number of parameters in the hypothesis.