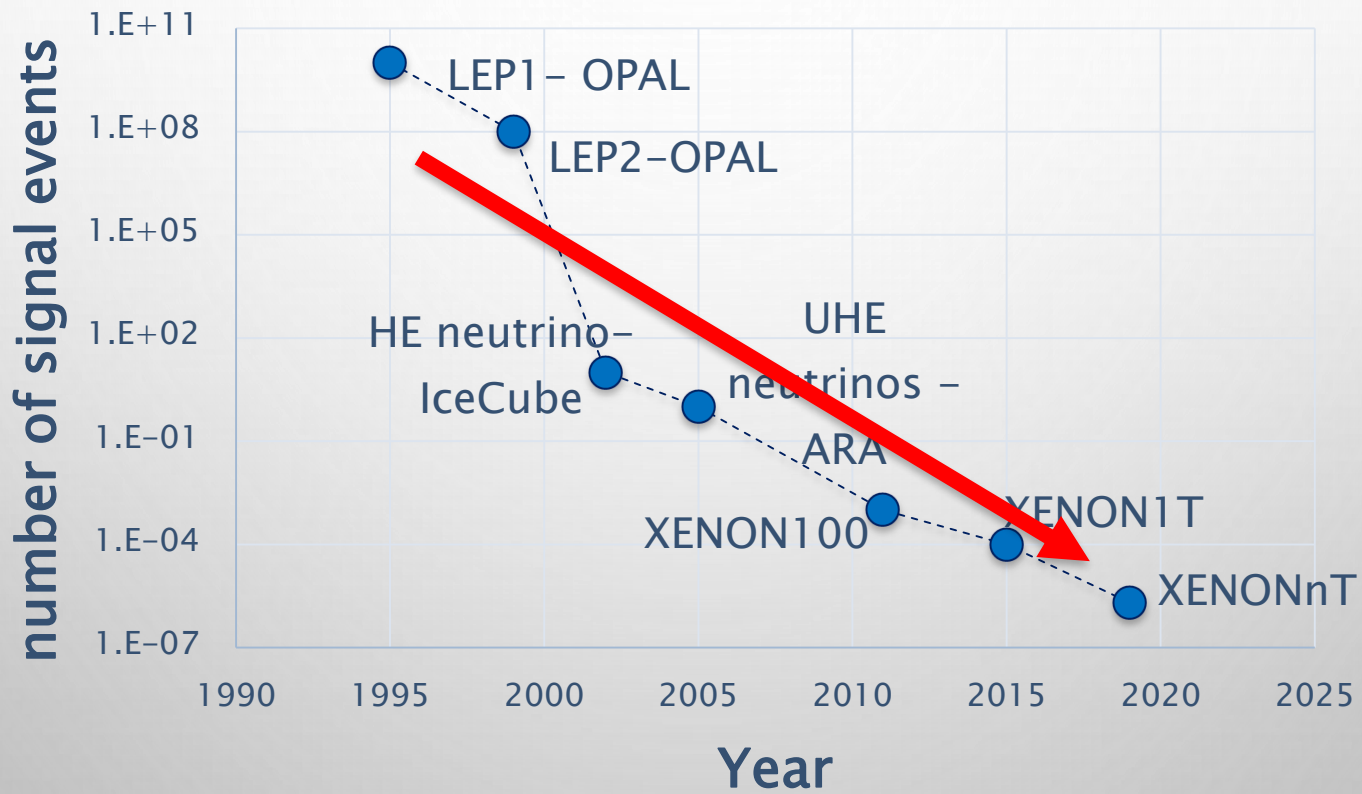


# LIMIT SETTING: EVOLUTION VS. INTELLIGENT DESIGN



Hagar Landsman  
Weizmann Institute of science

# MY CAREER



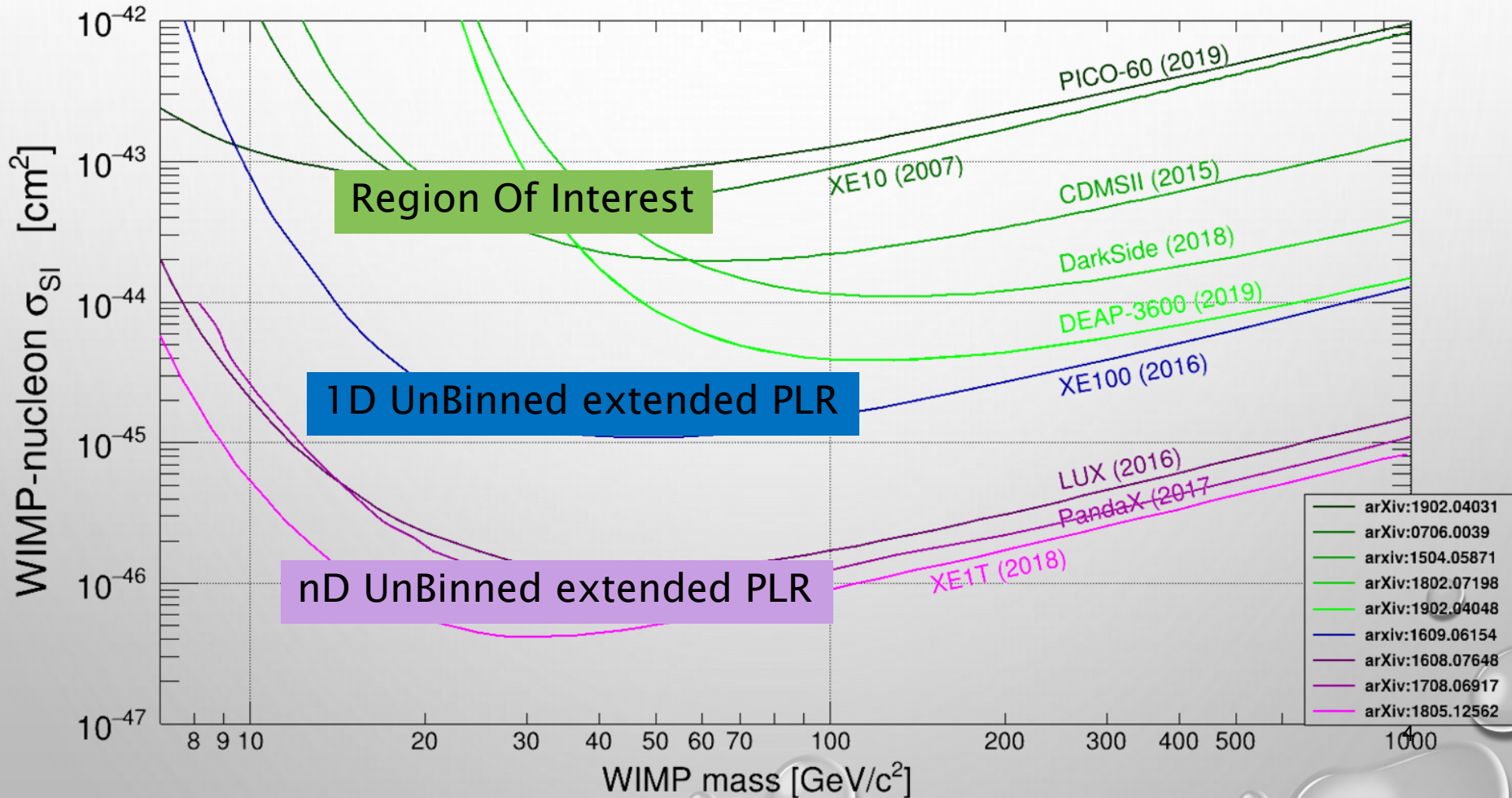
# THE EVOLUTION OF LIMITS SETTING

Region Of Interest

1D UnBinned extended PLR

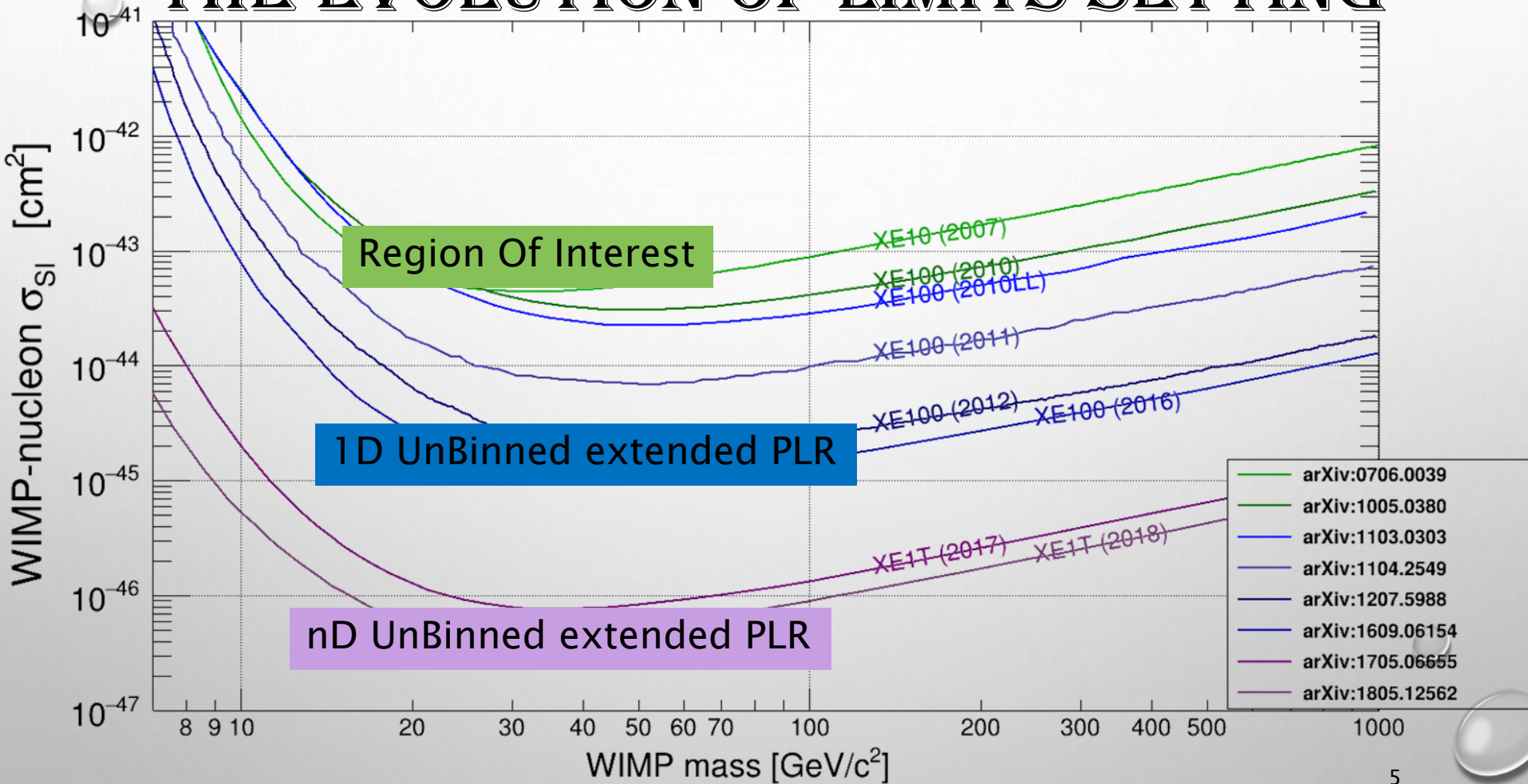
nD UnBinned extended PLR

# THE EVOLUTION OF LIMITS SETTING



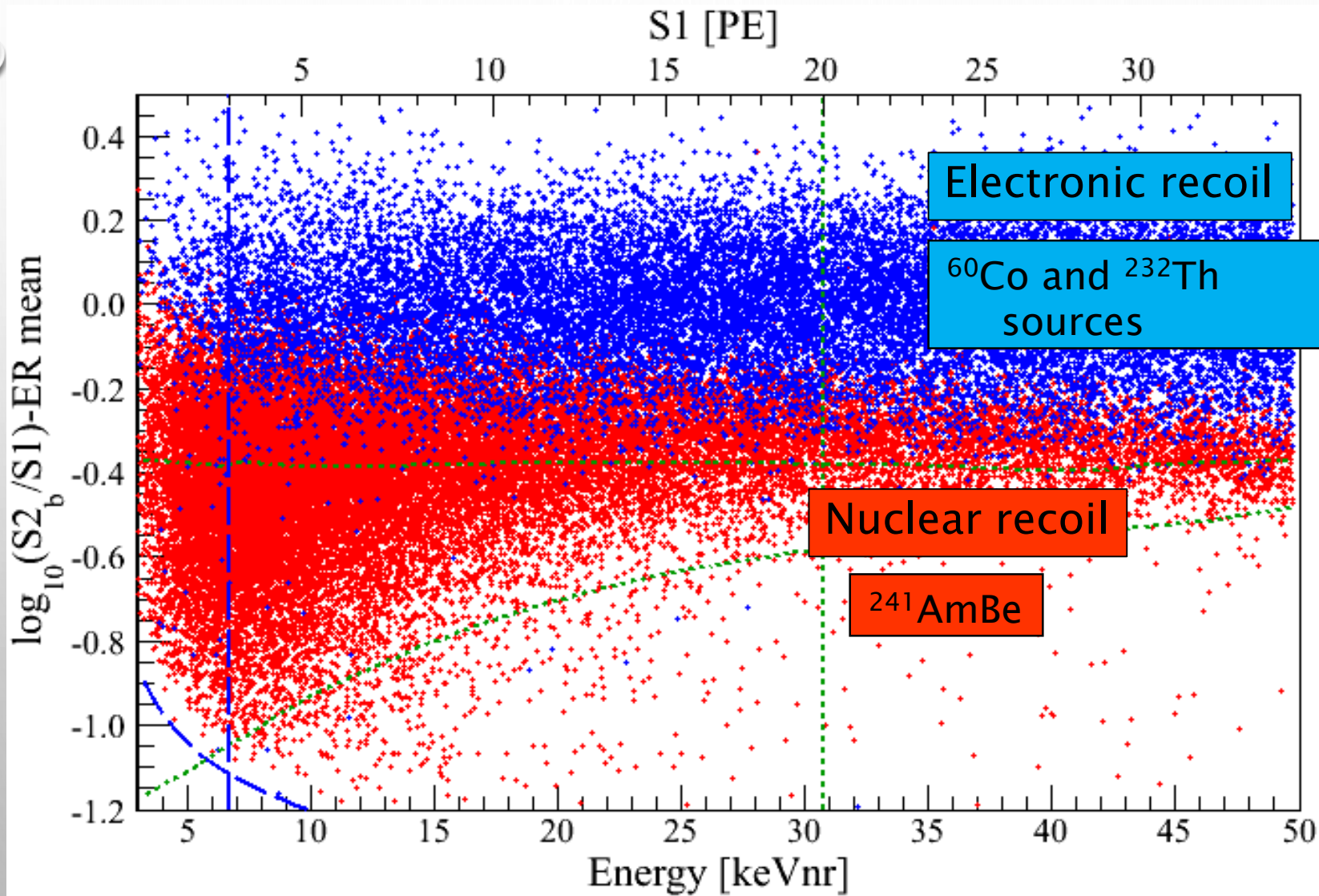
# XENON

## THE EVOLUTION OF LIMITS SETTING





# DETECTION PRINCIPLE DISCRIMINATION VARIABLES

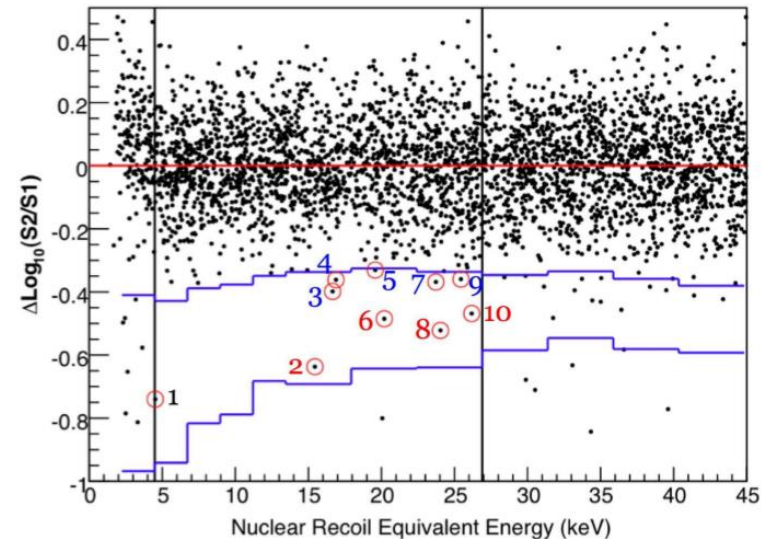
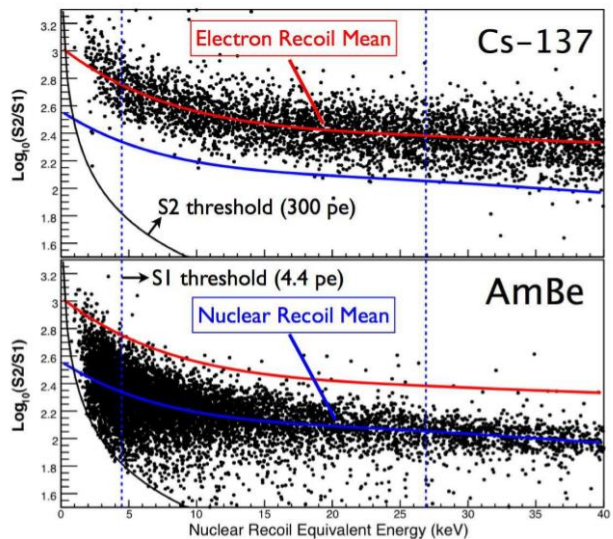


## Region Of Interest

First Results from the XENON10 Dark Matter Experiment  
at the Gran Sasso National Laboratory

J. Angle,<sup>1,2</sup> E. Aprile,<sup>3,\*</sup> F. Arneodo,<sup>4</sup> L. Baudis,<sup>2</sup> A. Bernstein,<sup>5</sup> A. Bolozdynya,<sup>6</sup> P. Brusov,<sup>6</sup> L. C. C. E. Dahl,<sup>6,8</sup> L. DeViveiros,<sup>9</sup> A. D. Ferella,<sup>2,4</sup> L. M. P. Fernandes,<sup>7</sup> S. Fiorucci,<sup>9</sup> R. J. Gaitskell,<sup>9</sup> K. R. Gomez,<sup>10</sup> R. Hasty,<sup>11</sup> L. Kastens,<sup>11</sup> J. Kwong,<sup>6,8</sup> J. A. M. Lopes,<sup>7</sup> N. Madden,<sup>5</sup> A. Manalaysay,<sup>1,2</sup> D. N. McKinsey,<sup>11</sup> M. E. Monzani,<sup>3</sup> K. Ni,<sup>11</sup> U. Oberlack,<sup>10</sup> J. Orboeck,<sup>2</sup> G. Plante,<sup>3</sup> R. Santorelli,<sup>3</sup> J. M. P. Shagin,<sup>10</sup> T. Shutt,<sup>6</sup> P. Sorensen,<sup>9</sup> S. Schulte,<sup>2</sup> C. Winant,<sup>5</sup> and M. Yamashita<sup>3</sup>

## XENON10 2005–2007

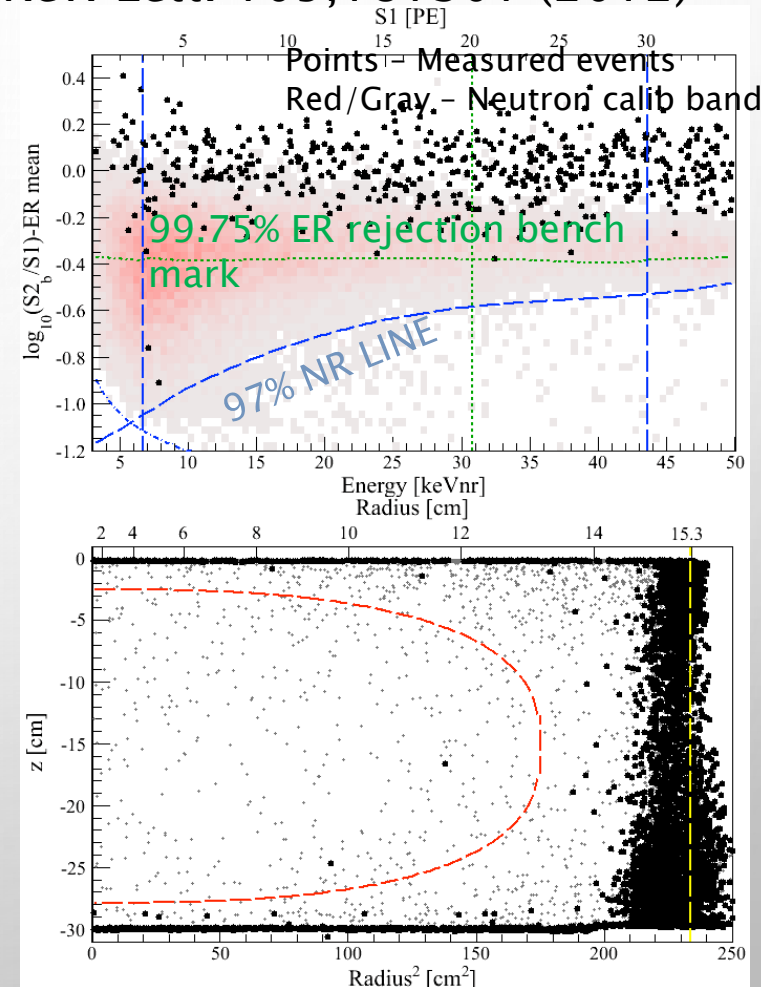


*“However, the uncertainty of the estimated number of leakage events for each energy bin in the analysis of the WIMP search data is currently limited by available calibration statistics. Based on the analysis of multiple scatter events, no neutron induced recoil event is expected in the single scatter WIMP-search data set. To set conservative limits on WIMP-nucleon spin-independent cross section, we consider all ten observed events, with no background subtraction.”*<sup>z</sup>

# WHAT DO WE LEARN FROM A TPC EVENT?

Illustrated by XENON100 2011/2012 data set  
 225 Live days Phys. Rev. Lett. 109,181301 (2012)

- $s_1, s_2$ :
  - Energy scale
  - Discrimination: ER vs. Nr ( $s_1/s_2$ )
- Vertex reconstruction
  - Fiducialization
  - Single vs. Multiple scatters
- Waveforms
- Event epoch time
- Slow control (detector stability)

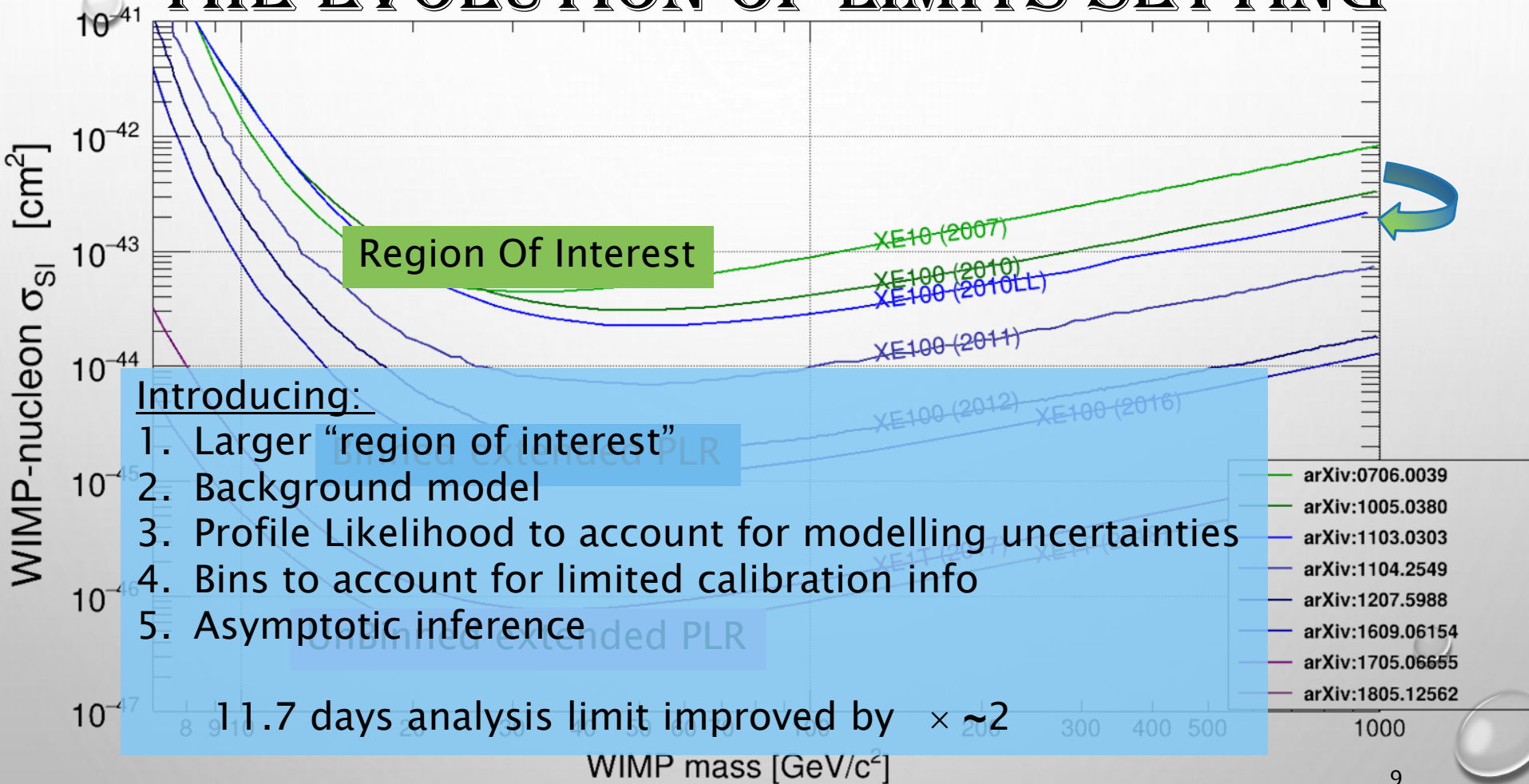


Bold – below 99.75 ER line (likely to be NR)  
 dots – above 99.75 ER line (likely to be ER)



# XENON

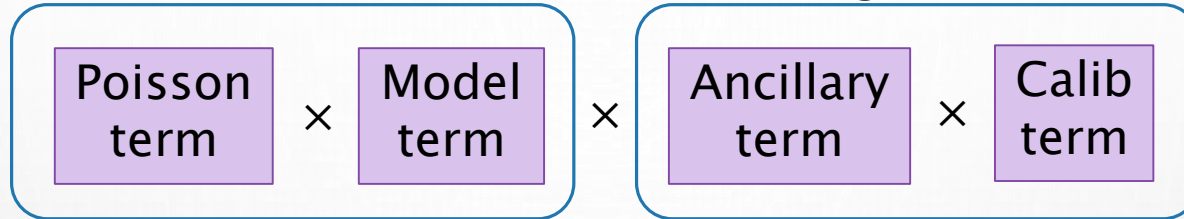
## THE EVOLUTION OF LIMITS SETTING



# THE LIKELIHOOD FUNCTION

Extended unbinned  $\mathcal{L}$

The guardians



$$\mathcal{L}_{ub} = \text{Poiss}(N|N_s + N_b) \prod_{i=1}^N \frac{N_s f_s(x_i) + N_b f_b(x_i)}{N_s + N_b}$$

$$N_s (\sigma; \bar{\theta}_s, \bar{\theta}_g)$$

$$N_b^j (\bar{\theta}_b, \bar{\theta}_g)$$

$n_{\text{data}}$

$$f_s (\bar{x} | \bar{\theta}_s, \bar{\theta}_g)$$

$$f_b^j (\bar{x} | \bar{\theta}_b, \bar{\theta}_g)$$

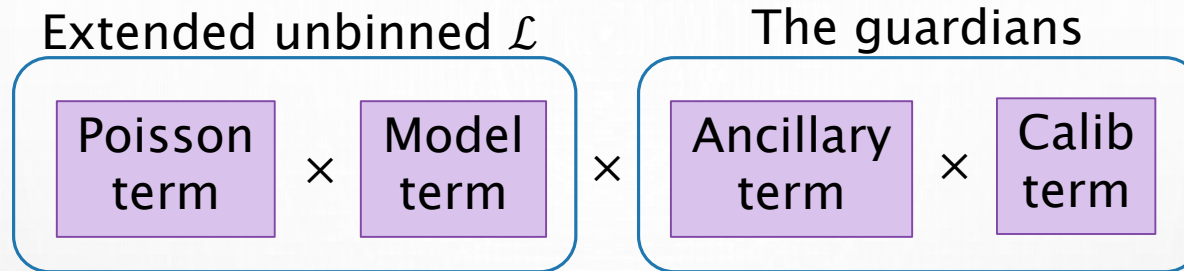
e. g.

$$\text{Gaus} \left( \hat{N}_b^j \mid N_b^j, \sigma^j \right)$$

$$\text{Gaus} \left( \hat{\theta} \mid \theta, \sigma_\theta \right)$$

- Long list of observables  $x$ :  $S_1, S_2, (R, z, \theta), t$
- Long list of parameters:  $\bar{\theta}_s, \bar{\theta}_g, \bar{\theta}_b$   
Some are correlated, some are not...

# THE LIKELIHOOD FUNCTION

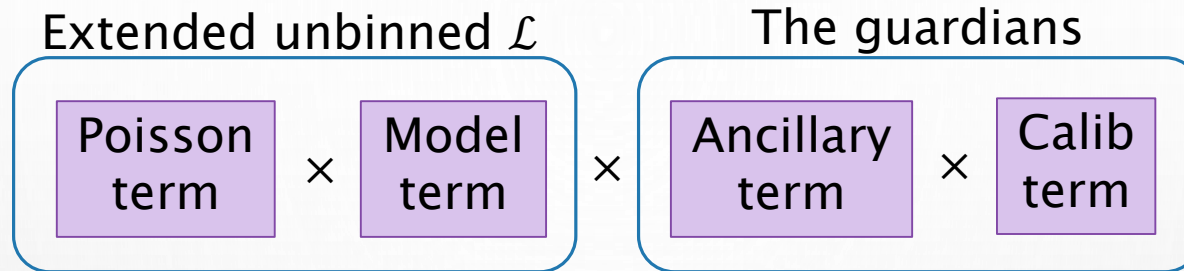


- Long list of observables  $x$ :  $S_1, S_2, (R, z, \theta), t$
- Long list of parameters:  $\bar{\theta}_s, \bar{\theta}_g, \bar{\theta}_b$

Three choices:

1. Ignore –  
That's easy to implement
2. Binned model –  
Bins in discrimination space (“bands”)  
Spatial bins ( $r, z, q$ )  
Temporal bins (E variations, background conditions, “runs”)  
$$\mathcal{L} = \mathcal{L}^I \times \mathcal{L}^{II} \times \mathcal{L}^{III}$$
3. Unbinned Model  
Higher dimensions for  $f_s, f_b$   
Add nuisance parameters

# THE LIKELIHOOD FUNCTION



Some (hopefully) good reasons to take it slowly:

→ Limited knowledge – risk of under/over coverage

- Limited calibration
- Lack of model
- Always risk of mis-modeling

→ Not needed

- The additional information / resolution is not needed

→ Save on resources

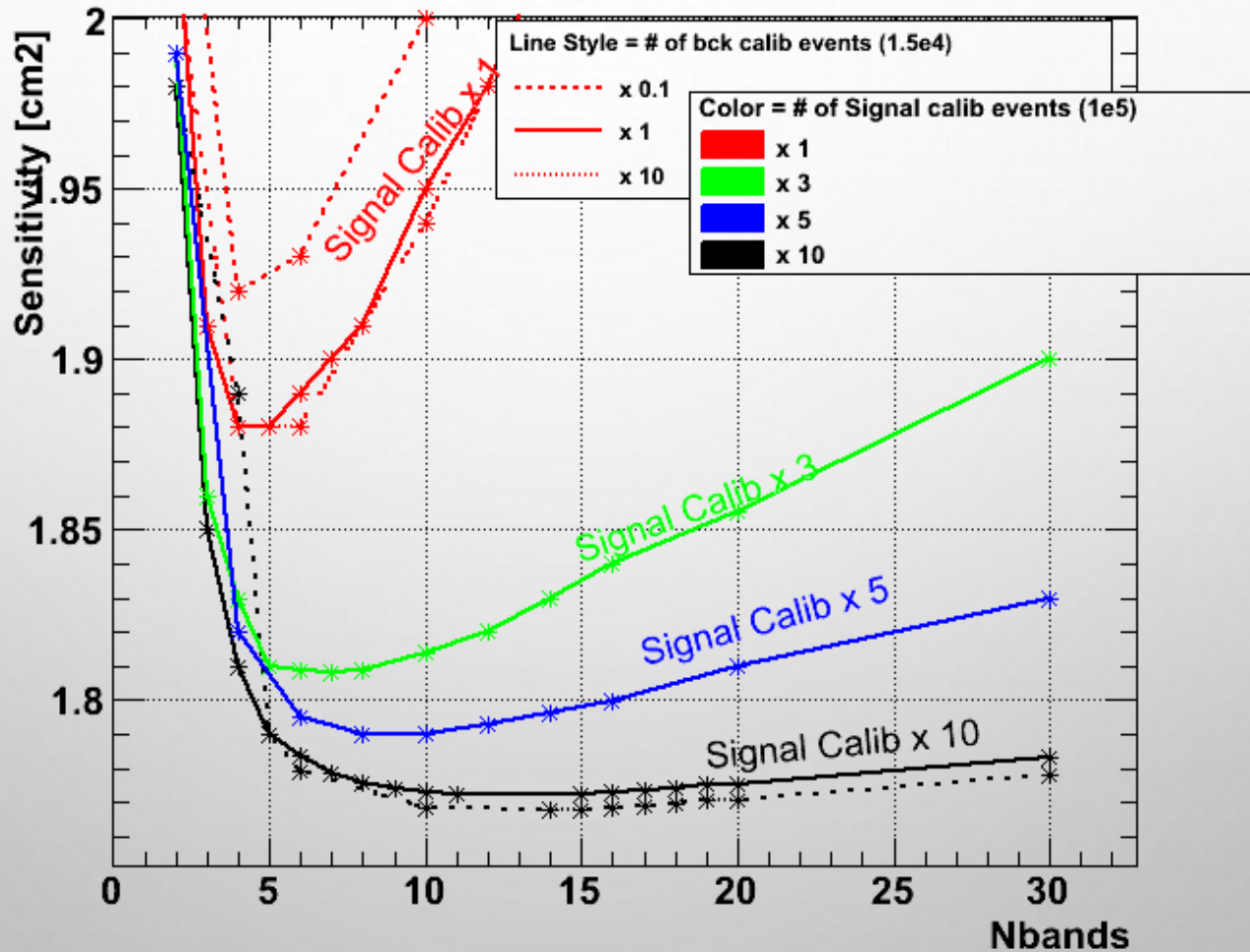
- Modeling and minimizing. Asymptoticness (checks or bypass)  
Required cpus, disk space, people, nerves, sanity

Use the power  
of the  
guardians  
wisely!



# HOW MANY BINS TO USE?

Sensitivity vs. Nbands, various Calib sizes (50 GeV, run 10)



# XENON'S 1<sup>ST</sup> LIKELIHOOD FUNCTION

Parameter of interest:

$N_s$  – total number of signal events

Nuisance parameters:

$N_b$  – background events

$\epsilon_s^i, \epsilon_b^i$  – distribution along bands of sig/bck

$t_{\text{Leff}}$  – deviation of  $L_{\text{eff}}$  from median

$$\mathcal{L} = \prod_j^K \text{Pois}(n^j | \epsilon_s^j N_s + \epsilon_b^j N_b) \quad \left. \vphantom{\prod_j^K} \right\} \text{Poisson on data, per band.}$$

$$\times \prod_{i=1}^{n^j} \frac{\epsilon_s^j N_s f_s(S1_i) + \epsilon_b^j N_b f_b(S1_i)}{\epsilon_s^j N_s + \epsilon_b^j N_b} \quad \left. \vphantom{\prod_{i=1}^{n^j}} \right\} \text{Distribution of events in each band}$$

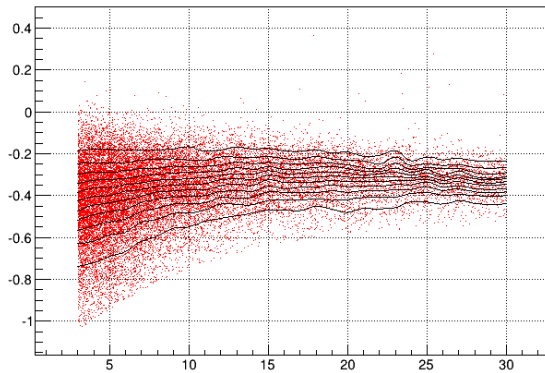
$$\times \prod_j^K \text{Pois}(m_s^j | \epsilon_s^j M_s) \times \prod_j^K \text{Pois}(m_b^j | \epsilon_b^j M_b) \quad \left. \vphantom{\prod_j^K} \right\} \text{Poisson on calibration data}$$

$$\times \exp(-(t - t_{\text{obs}})^2 / 2) \quad \left. \vphantom{\exp} \right\} \text{Leff penalty}$$

# XENON100 - PLR RESULTS

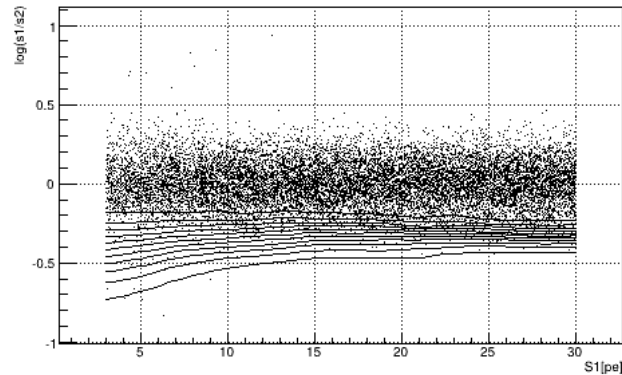
Using AmB data to construct bands.  
Cross check using MC distribution

data/run10/ambe\_34kg.dat



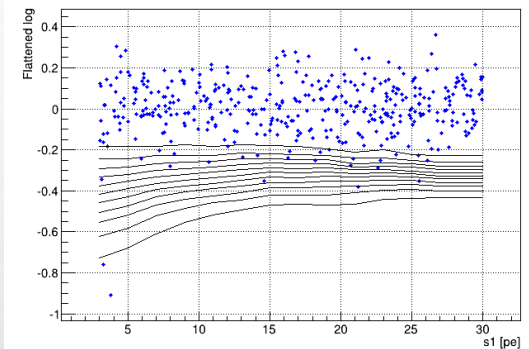
Nuclear Recoils

ER data for run10



Electronic Recoils

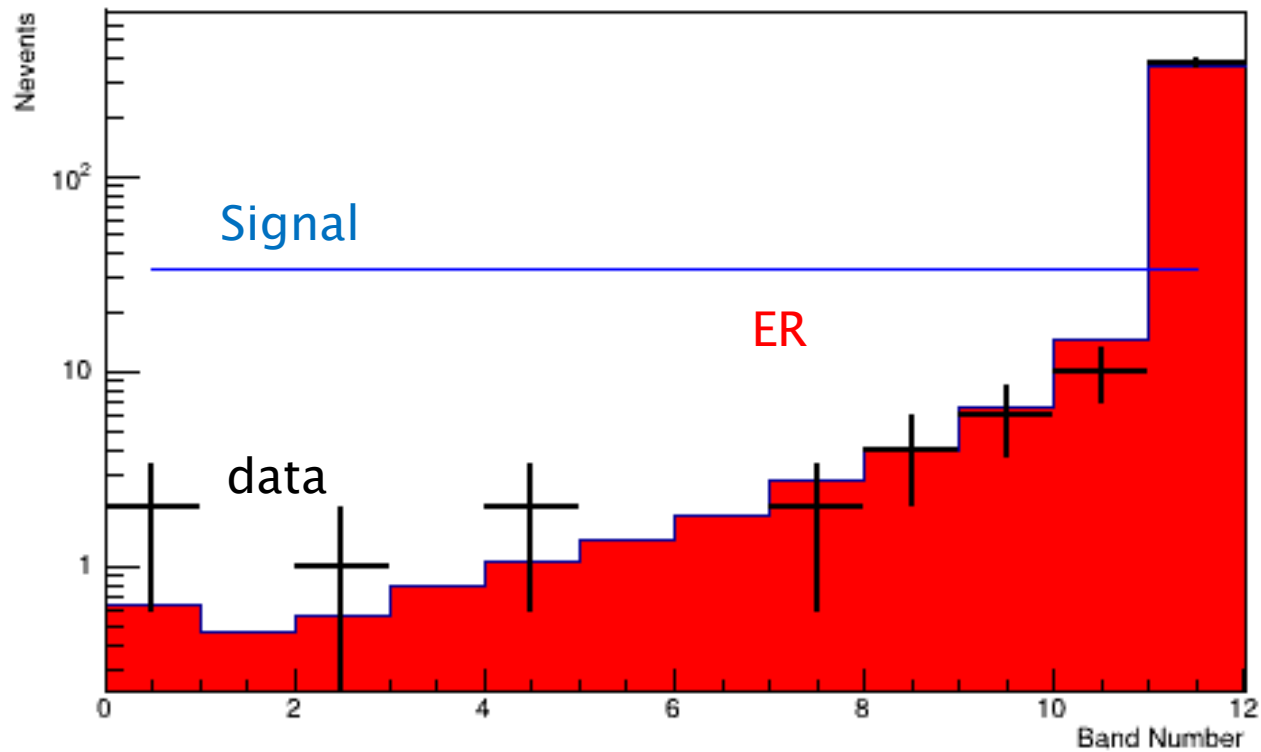
Run 10 data set



Data

# XENON100 - PLR RESULTS

Distribution between bands (ER-red,NR-blue,data-black)

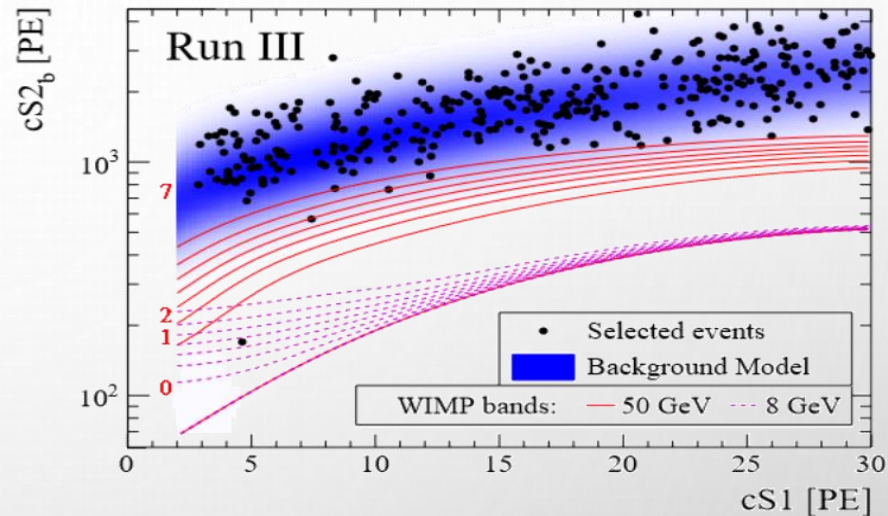




# XENON100 - PLR RESULTS- COMBINATION 2016 ANALYSIS

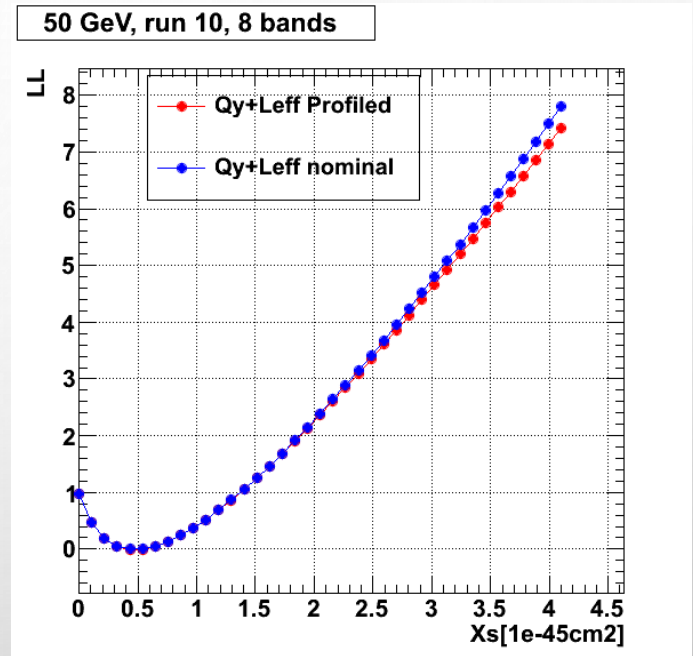
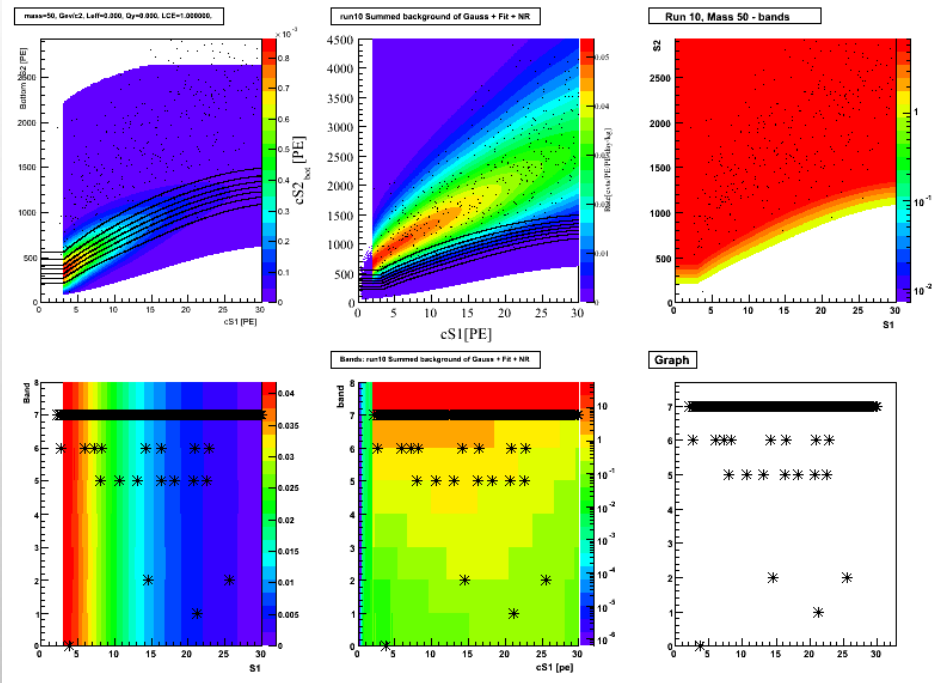
## Introducing...

- Use MC for S2 modeling (instead of AmBe cal)
- Different bands for different masses
- Additional model nuisance parameter  $L_{\text{eff}}$  and  $Q_y$

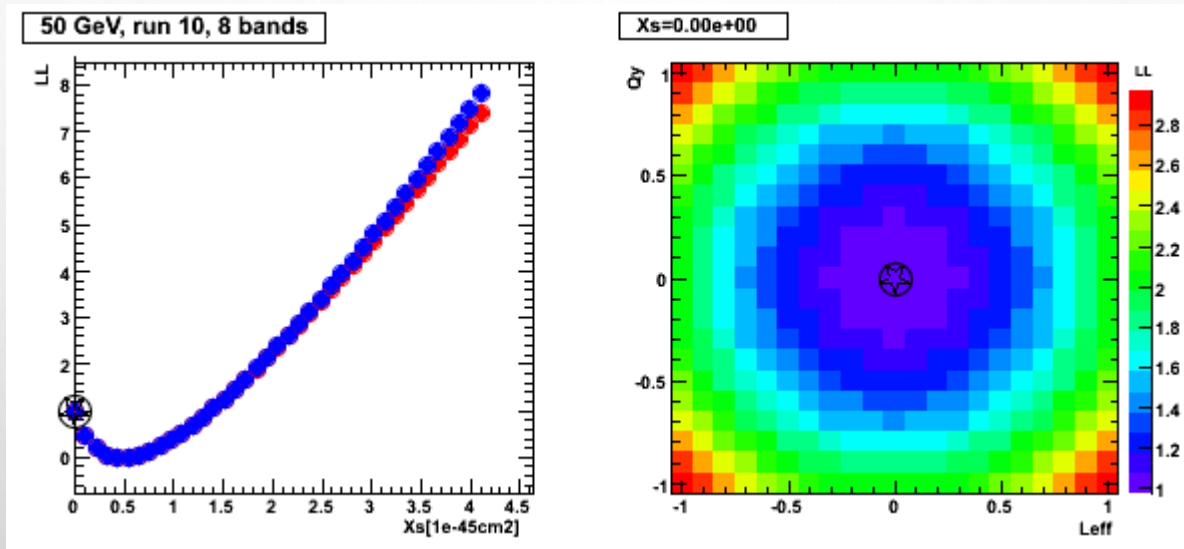


- Divide and conquer – PDF simulation production was done offline...and not during minimization
- Two variables: S1, and cS1 (and not just cS1)  
To better account for light collection efficiency variations

# 225 DAYS, 50 GEV

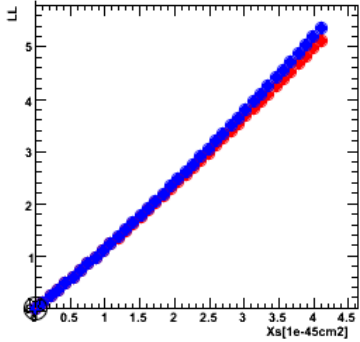


# 225 DAYS, 50 GEV

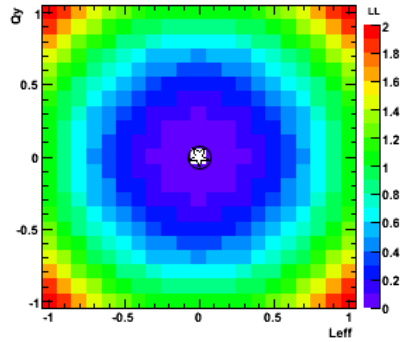


# RUN COMBINATION

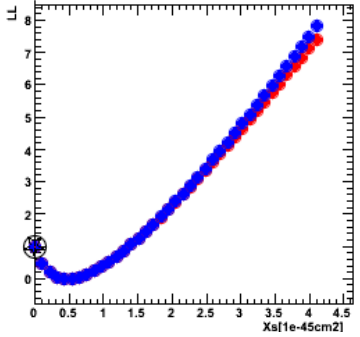
50 GeV, run 8, 8 bands



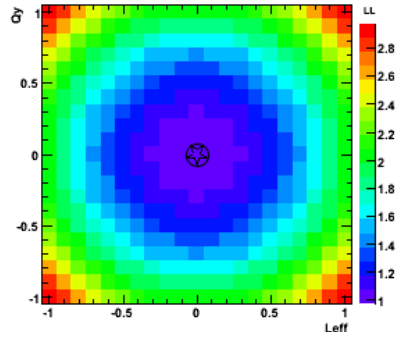
$X_s=0.00e+00$



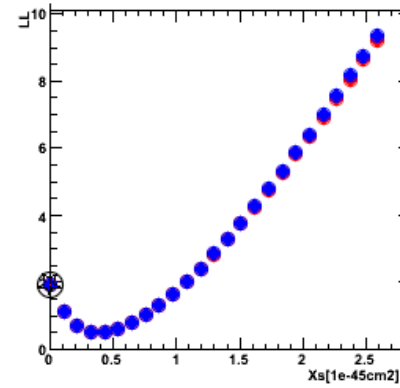
50 GeV, run 10, 8 bands



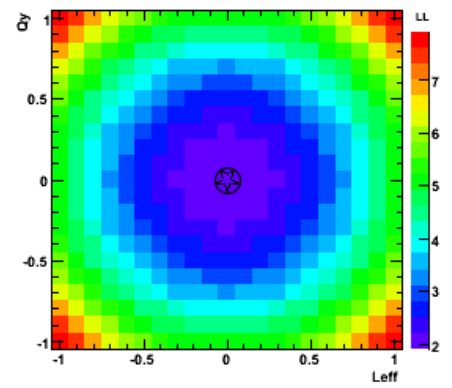
$X_s=0.00e+00$



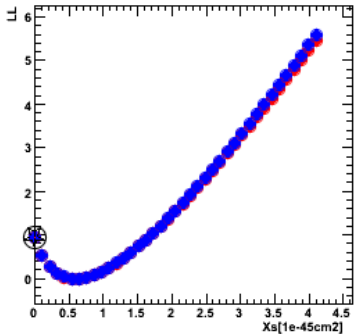
50 GeV, run 30, 8 bands



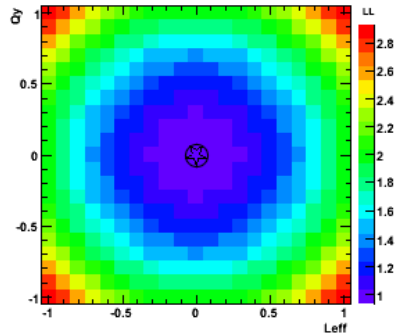
$X_s=0.00e+00$



50 GeV, run 12, 8 bands



$X_s=0.00e+00$

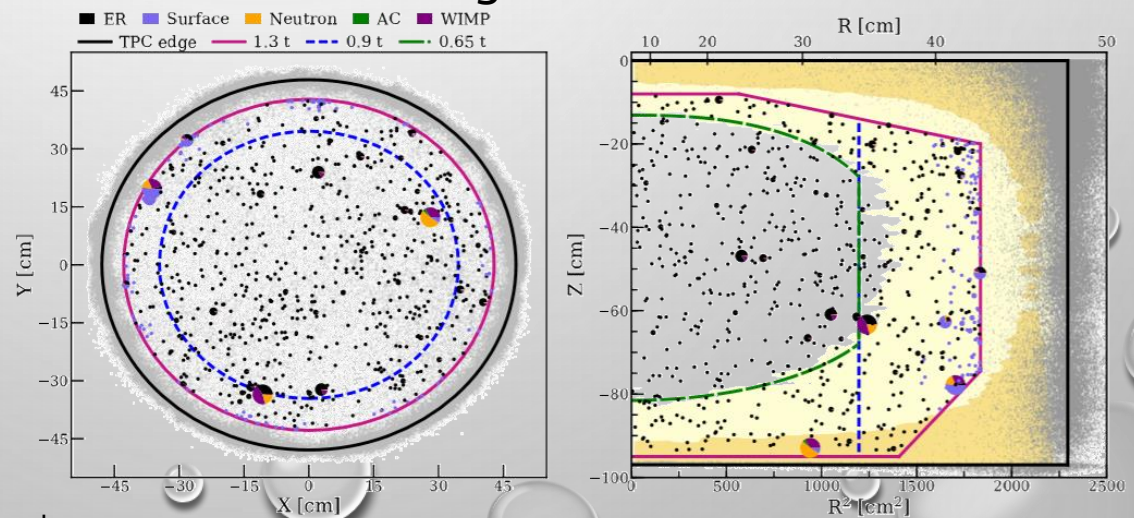




# XENON1T - PLR RESULTS- 1TON-YEAR 2018 ANALYSIS

## Introducing...

- PDFs in higher dimensions ( $s_1, s_2, r$ ). No bands in  $s_2$ .
- Larger volume used
- 4 independent background models constraint by calibration and simulation
- More nuisance parameters
- More complete interaction model
- More sophisticated background model with some a-priori fits
- Safeguard to account for some mis-modeling

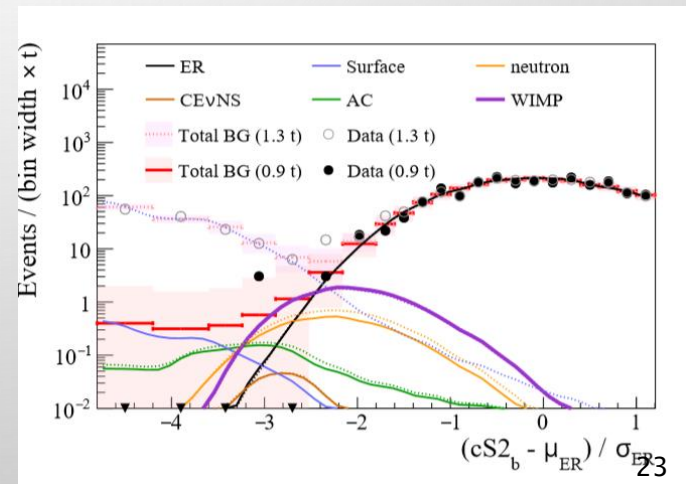
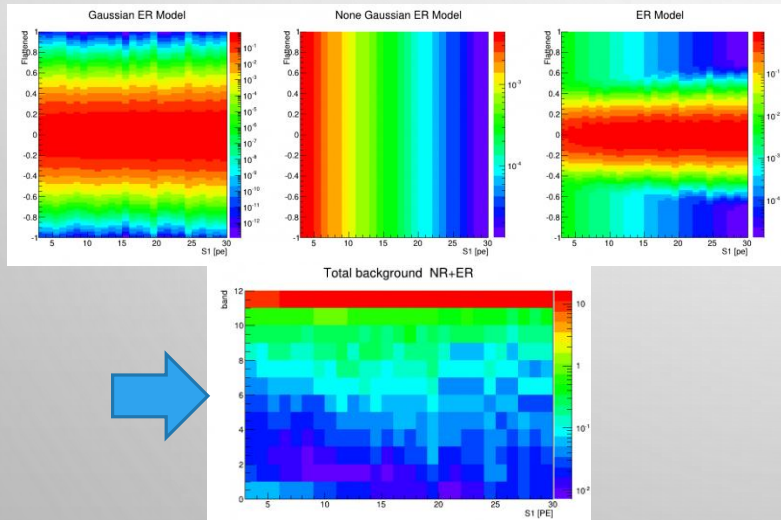


# SOME THOUGHT ON SIGNAL MODEL

- Signal model sets:  $f_s$  and  $n_s(\sigma)$  and  $(\varepsilon_s)$
- Don't forget our parameter of interest is  $\sigma$  ( not  $n_s$ )
- Energy scale:  $\text{pe} \leftrightarrow \text{kev}_{\text{nr}}$
- Nuisance parameters in astrophysical model, interaction model, detector response
- No calibration sample available  
(calibration data can be used to constraint parameters)
- Need to artificially incorporate spatial and temporal detector instabilities

# SOME THOUGHT ON BACKGROUND MODEL

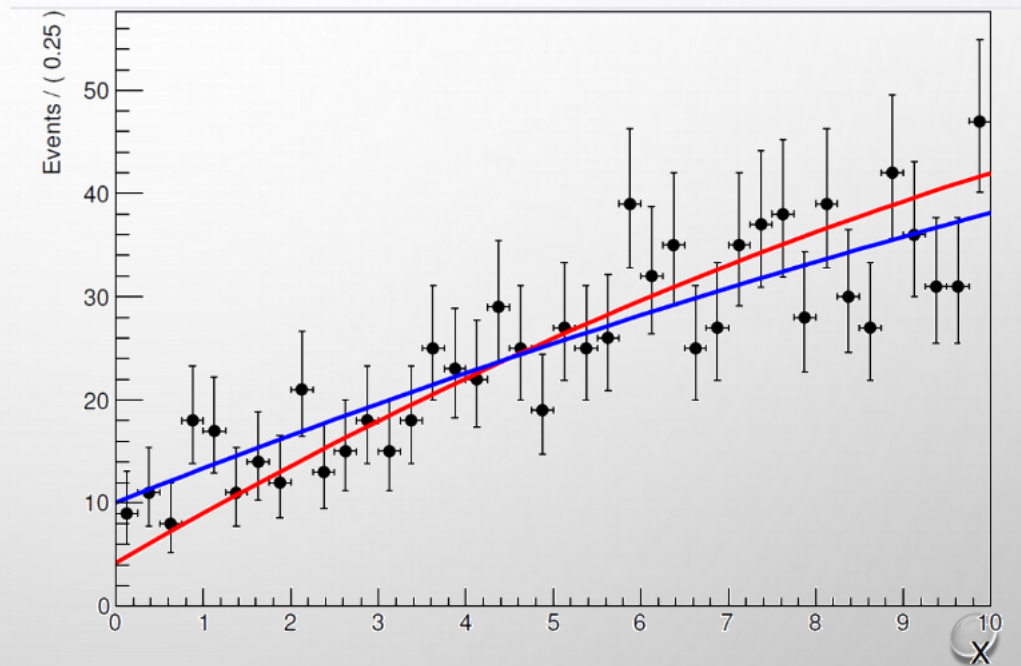
- Background model sets:  $f_b$  ( $\varepsilon_b$ ) and sometimes  $N_b$
- Several components of background: Fractions can be “frozen” or be nuisance
- Shape and magnitudes modeling
- Calibration samples may exist – statistic decreases with #variables
- Is our background model accurate “enough”?



# THE CURSE OF BACKGROUND MISMODELLING

## THE PROBLEM

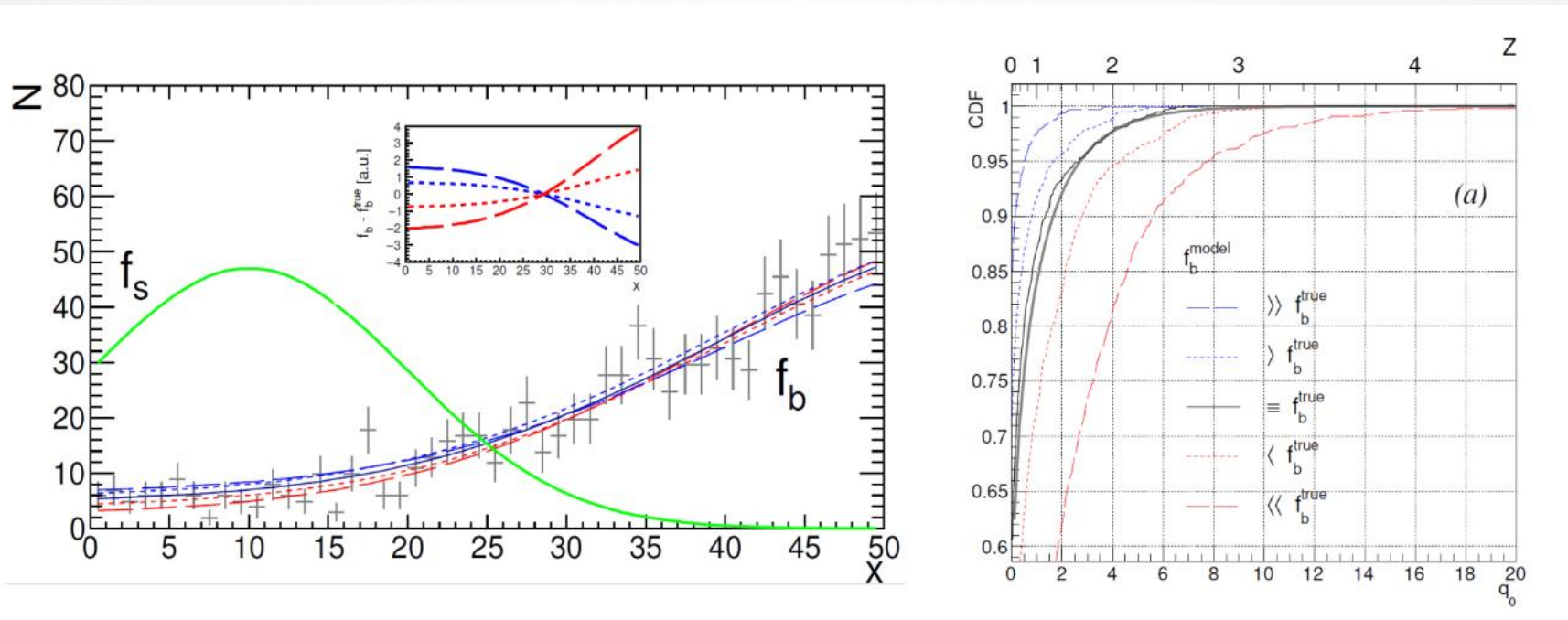
- Too many parameters
- Hidden parameters
- Partial underlying model
- ....Mistakes...



Might lead to enhanced false discovery rate or overly constrained limits



# THE CURSE OF BACKGROUND MISMODELLING THE PROBLEM



Arxiv:1610.02643



# THE CURSE OF BACKGROUND MISMODELLING

## THE PROBLEM

Arxiv:1610.02643

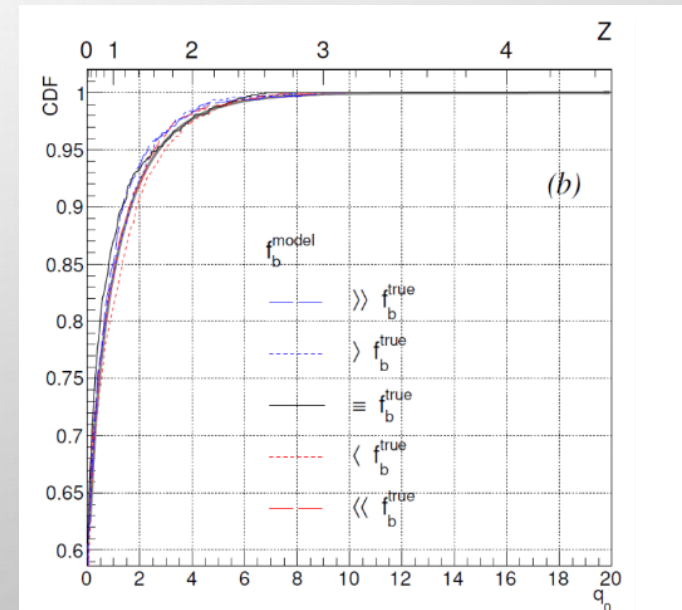
- Use the benchmark model
- Do not add extra nuisance parameters

$$f_b(x) \rightarrow (1 - \varepsilon) f_b(x) + \varepsilon f_s(x)$$

$$L_{overall} = Poiss(N|N_s + N_b) \prod \frac{N_s f_s(x_i) + N_b (1 - \varepsilon) f_b(x_i) + N_b \varepsilon f_s(x_i)}{N_s + N_b} \times L_{cal}(\varepsilon)$$

$$L_{cal}(\varepsilon) = \prod (1 - \varepsilon) f_b(x_i) + \varepsilon f_s(x_i)$$

- Works for limits and discoveries
- Safeguards background components that are based on calibration
- We found out that a similar technique used for cross checks in the LHC, “spurious signal”



# WHERE IT HURTS...

Many delicate points and challenges to address.

Here are just a few examples\*:



# WHERE IT HURTS...

## EXAMPLE 1: “THE CURSE OF MISMODELLING”

The “safeguard” can provide some protection for models constructed based on calibration samples.

Nuisance parameters can be added, but

- Require some model assumption
- Complicates analysis - heavier, slower

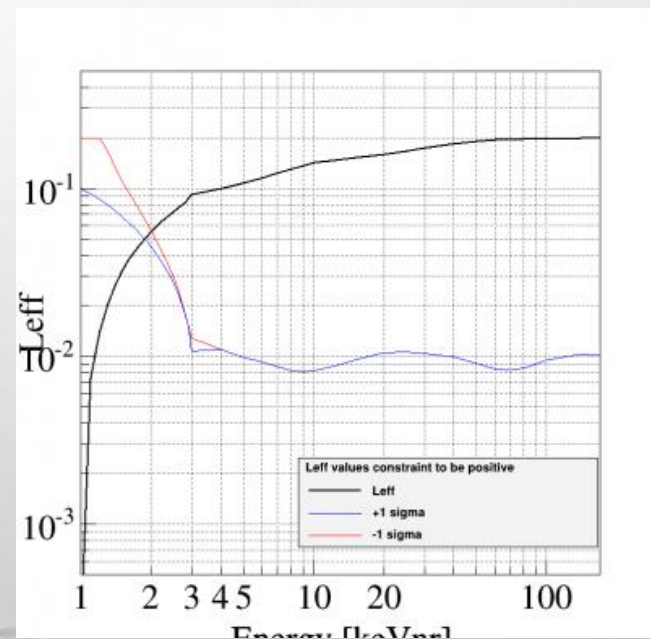
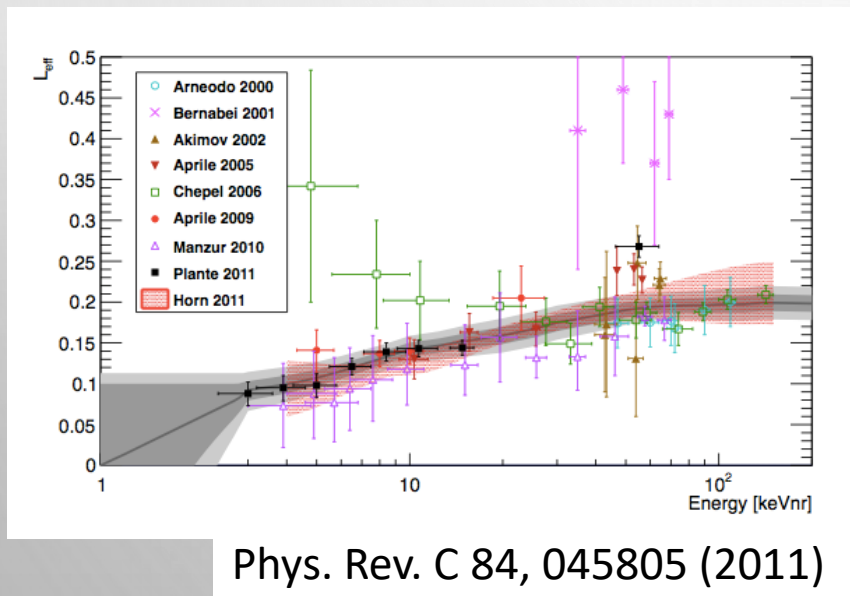
It is not enough

# WHERE IT HURTS...

## EXAMPLE 2: "THE CURSE OF THE UN-MODELLED"

- Include nuisance parameters without an underlying model
- Non physical regions
- Non symmetric nuisance uncertainties

E.G.  $L_{\text{eff}}$



# WHERE IT HURTS...

## EXAMPLE 3: "THE BLESSING OF ASYMPTOTICNESS"

(Or "we ♥ wilks & arxiv1007.1727")

Low bg

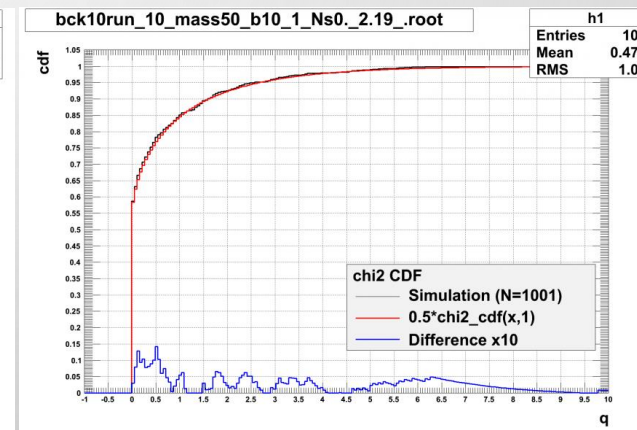
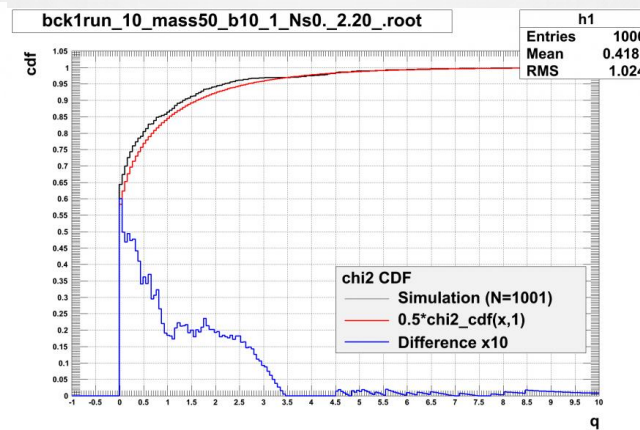
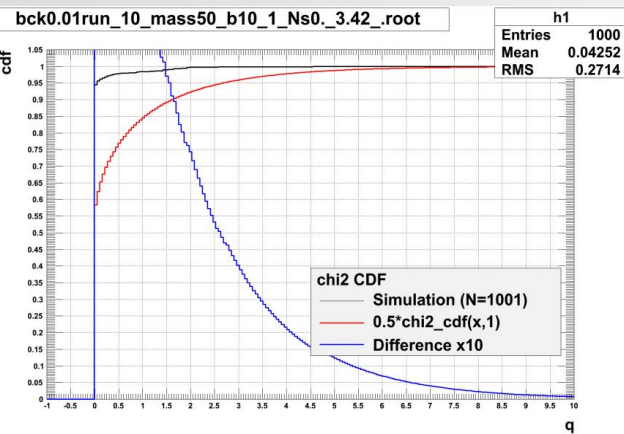
$$q_\sigma = \begin{cases} -2 \ln \lambda(\sigma) & \hat{\sigma} < \sigma \\ 0 & \hat{\sigma} > \sigma \end{cases}$$

$$p_s = \int_{q_\sigma^{\text{obs}}}^{\infty} f(q_\sigma | H_\sigma) dq_\sigma.$$

Ncalib=151

Ncalib=15128

Ncalib=151280



Need to verify asymptoticness and run MC if broken

30



# WHERE IT HURTS...

## EXAMPLE 4: “THE CURSE OF MULTIPLE DIMENSIONS”

Generating multidimensional ( $s_1, s_2, r, z, \dots$ ) pdf maps for “many” nuisance parameters variations

- Algorithm:

- Prepare a model bank ahead of time

- Or build the necessary model during minimization

- (Possibly with smart book keeping and archiving)

- Nuisance parameter resolution

- How large a step in modeling

- Interpolate?

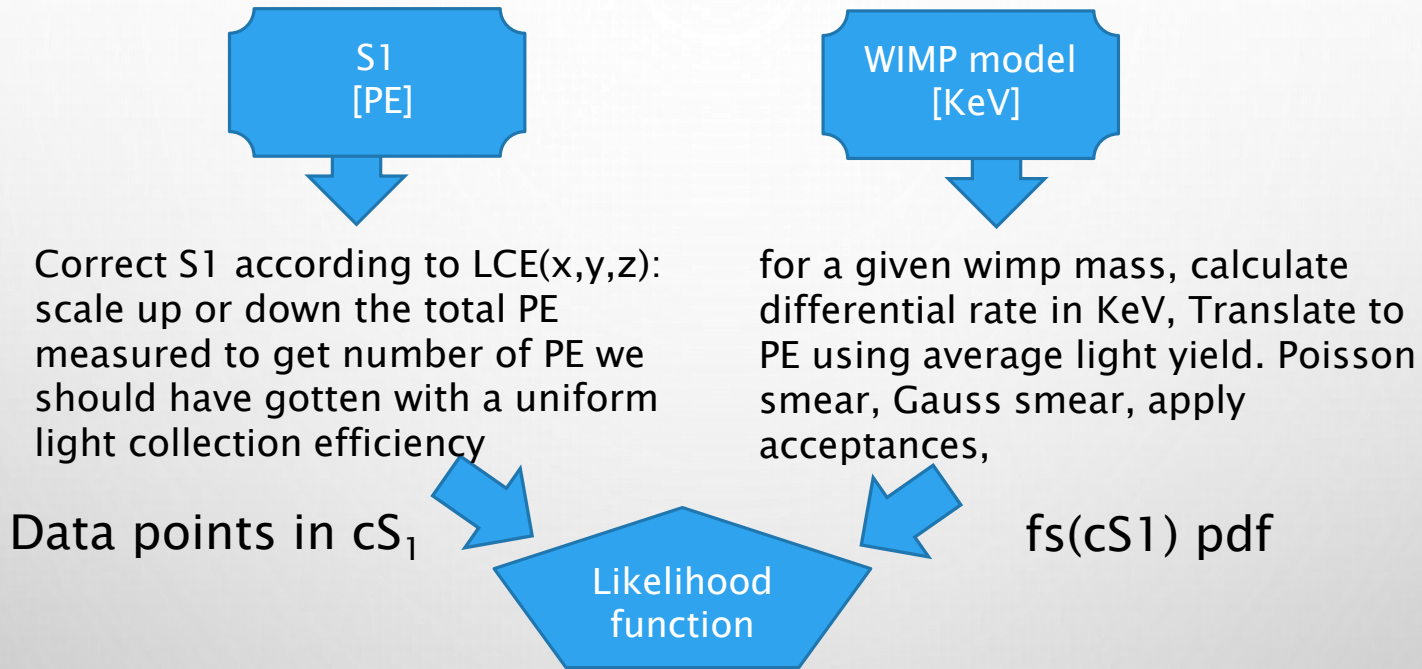
- Verifying asymptoticness or doing mc instead becomes painful

- Also: complicated codes

$$\begin{aligned}\lambda(\sigma) &= \frac{\max_{\sigma \text{ fixed}} \mathcal{L}(\sigma; \mathcal{L}_{\text{eff}}, v_{\text{esc}}, N_b, \epsilon_s, \epsilon_b)}{\max \mathcal{L}(\sigma, \mathcal{L}_{\text{eff}}, v_{\text{esc}}, N_b, \epsilon_s, \epsilon_b)} \\ &\equiv \frac{\mathcal{L}(\sigma, \hat{\mathcal{L}}_{\text{eff}}, \hat{v}_{\text{esc}}, \hat{N}_b, \hat{\epsilon}_s, \hat{\epsilon}_b)}{\mathcal{L}(\hat{\sigma}, \hat{\mathcal{L}}_{\text{eff}}, \hat{v}_{\text{esc}}, \hat{N}_b, \hat{\epsilon}_s, \hat{\epsilon}_b)}.\end{aligned}$$

# WHERE IT HURTS...

## EXAMPLE 5: "THE CURSE OF HANDWAVING"



Loop on all events in each band. For each event, use its  $cS_1$  to check how likely it is to come from the signal pdf, or background pdf.

**Problem:  $cS_1$  is not physical. Low PE cut, Poisson smearing should be done on  $s_1$ !**

# WHERE IT HURTS...

## EXAMPLE 6: "THE CURSE OF DIVERSITY"

$$p'_s = \frac{p_s}{1 - p_b}$$

### e.g. Over coverage:

- Power constraint
- CIs (Roughly 90%CL→95%CL)
- Ce la vie

where

$$1 - p_b = \int_{q_\sigma^{\text{obs}}}^{\infty} f(q_\sigma | H_0) dq_\sigma$$



# WHERE IT HURTS...

## EXAMPLE 7: “THE CURSE OF PAGE LIMIT”

- Many details to the models, inference method...
- Information in papers is limited. Very often summarized to: “...as was done in [xx].”
- Would be nice to see more detailed likelihood functions...
- Would be nice to see more likelihood curves...
- Many consistency checks, verifications to be made usually not even explicitly acknowledged.
- Follow up papers become more popular, but ...cannot make everyone happy....

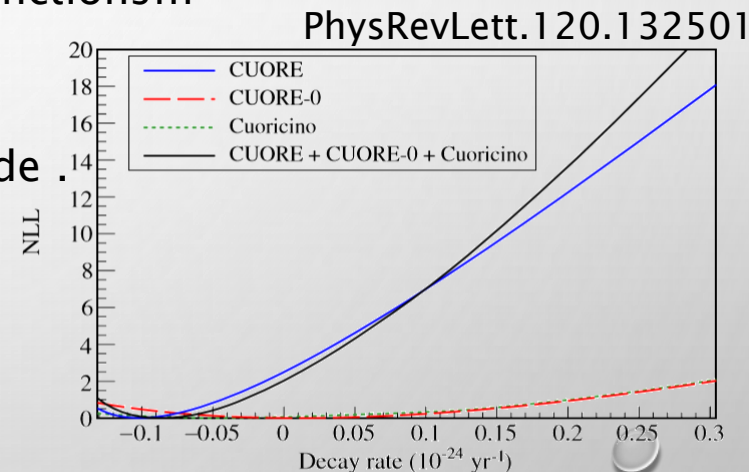
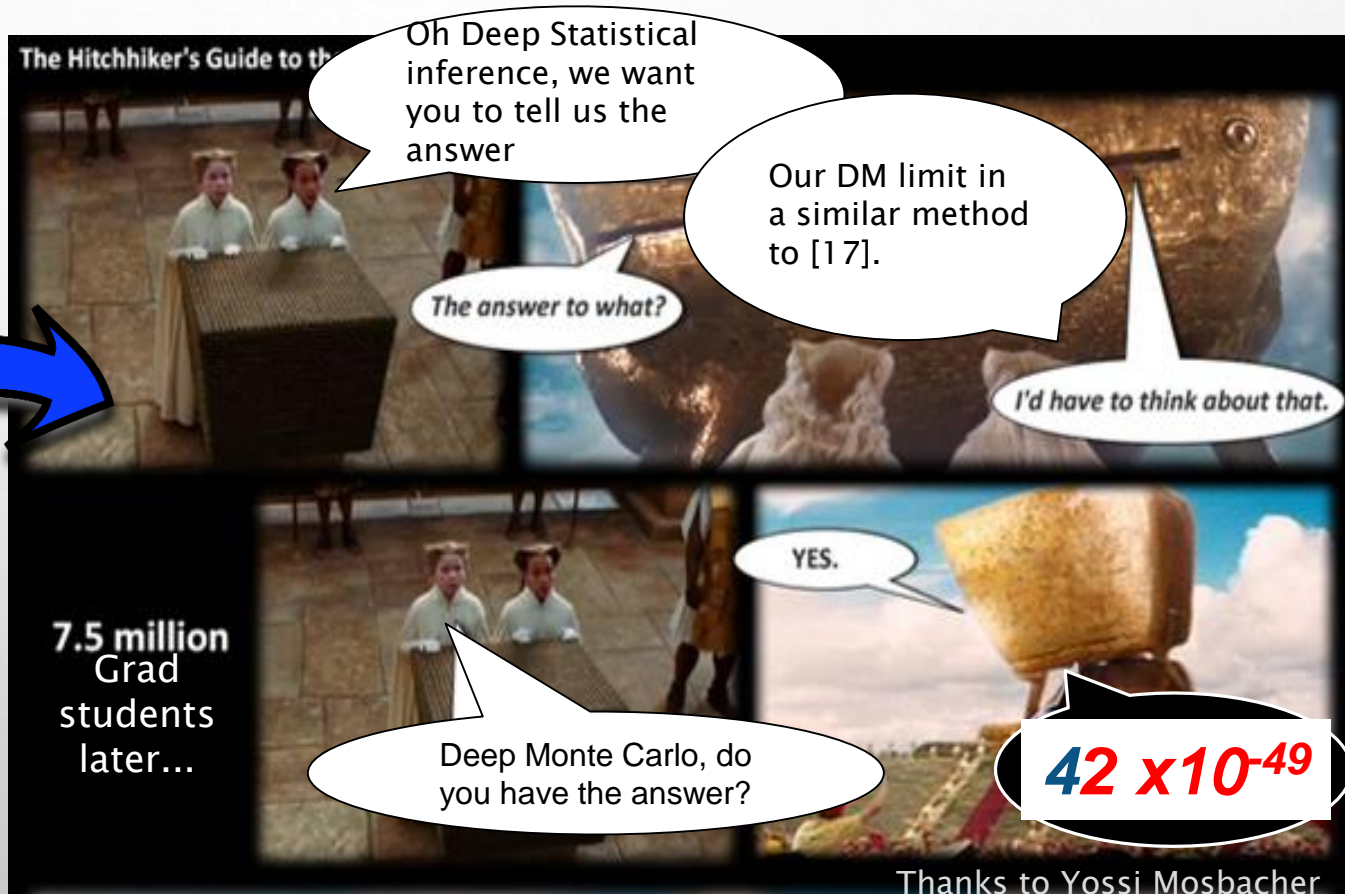
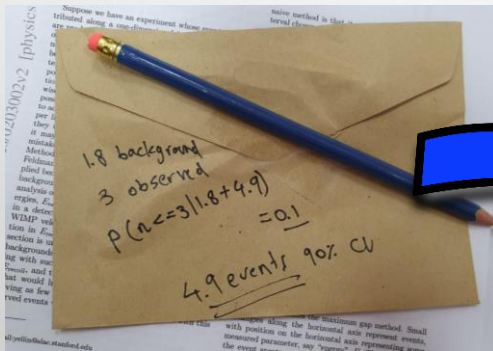


FIG. 4. Profile negative-log-likelihood curves for CUORE, CUORE-0, Cuoricino, and their combination.

# GRAPHICAL SUMMARY

Back then....



7.5 million  
Grad  
students  
later...

**$42 \times 10^{-49}$**

Thanks to Yossi Mosbacher