

# “Point 5”: LOI – DBD comparison

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ILD Phone Meeting – 24.09.2014

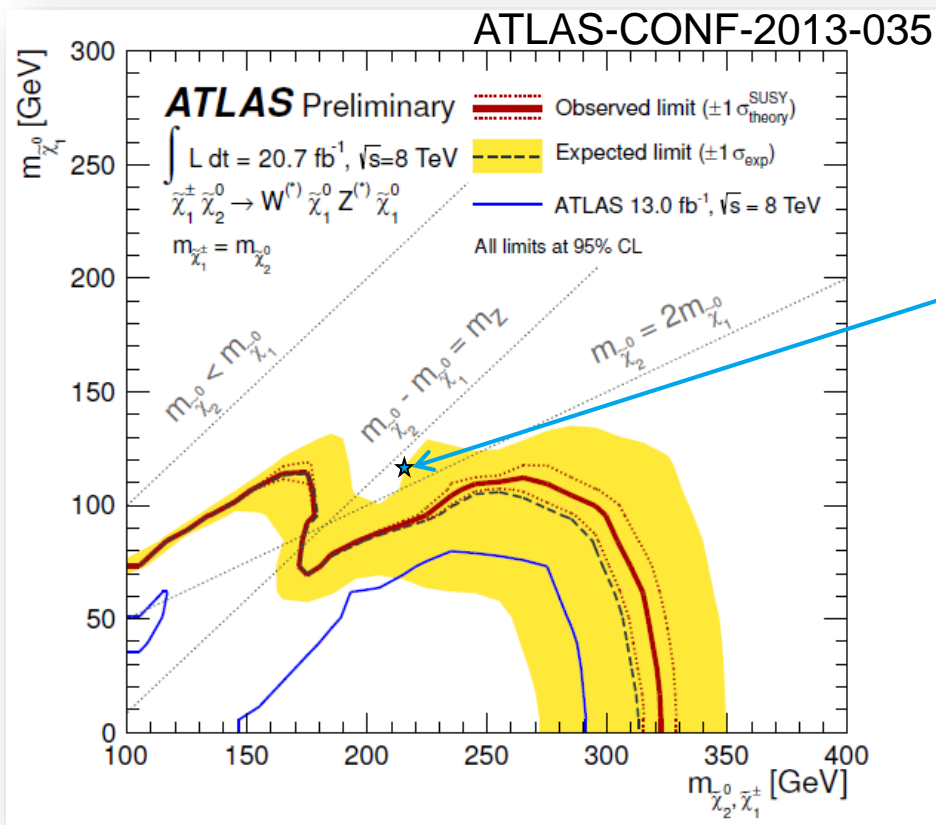


# Point 5: $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_2^0$ Pair Production at the ILC

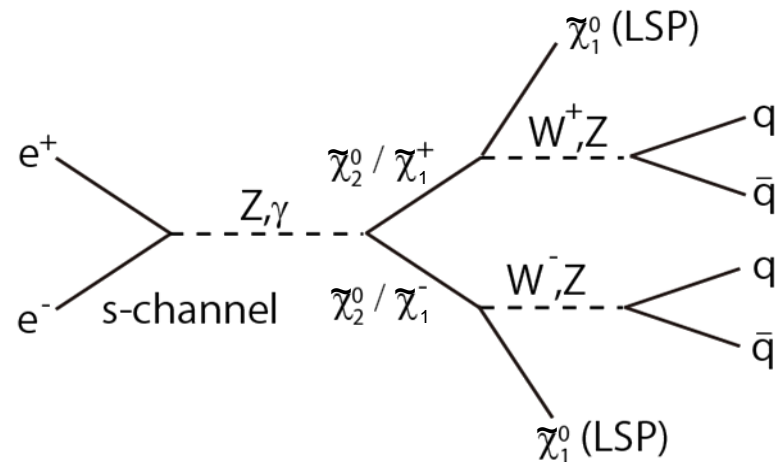
“Point 5” benchmark : gaugino pair production at ILC

<http://arxiv.org/pdf/1006.3396.pdf> (ILD Lol)

<http://arxiv.org/pdf/0911.0006v1.pdf> (SiD Lol)



| Particle             | Mass [GeV] |
|----------------------|------------|
| $\tilde{\chi}_1^0$   | 115.7      |
| $\tilde{\chi}_1^\pm$ | 216.5      |
| $\tilde{\chi}_2^0$   | 216.7      |
| $\tilde{\chi}_3^0$   | 380        |

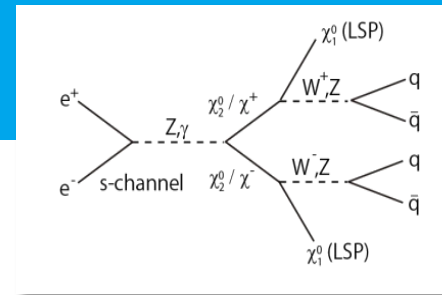


$$\tilde{\chi}_1^\pm \rightarrow \tilde{\chi}_1^0 W^\pm \quad BR = 99.4\%$$

$$\tilde{\chi}_2^0 \rightarrow \tilde{\chi}_1^0 Z^0 \quad BR = 96.4\%$$



# Study case - motivation

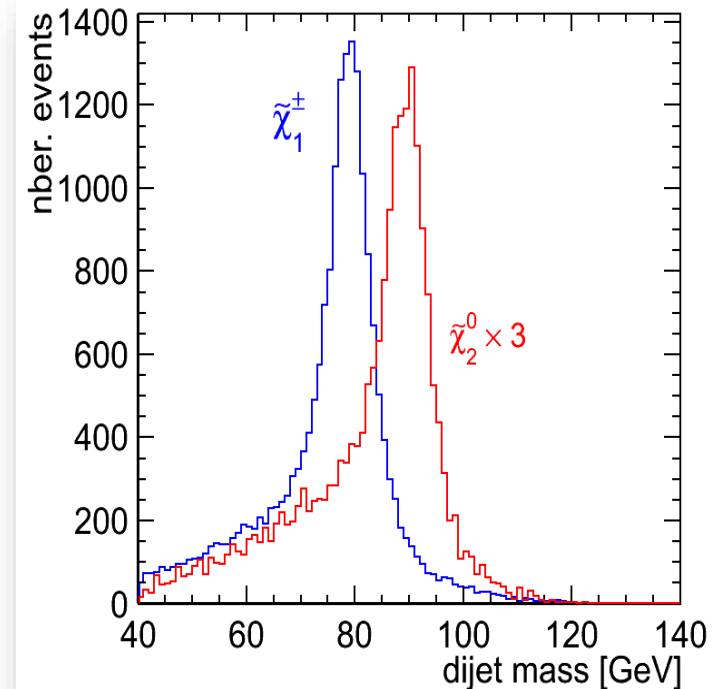


## ➤ Signal topology:

- **Four jets** and **missing energy** (due to LSP)
- **Hadronic decay** modes of gauge bosons chosen as **signal**
- Both decay channels treated as signal in turn

$$\tilde{\chi}_1^\pm \rightarrow \tilde{\chi}_1^0 W^\pm \quad \text{and} \quad \tilde{\chi}_2^0 \rightarrow \tilde{\chi}_1^0 Z^0$$

- $\tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  sample separation: essentially **distinguish between W and Z** pair events
- Good case for studying the detector and particle flow performance



# Data Samples:

> Signal: 40000  $\tilde{\chi}_1^\pm$  events and 9000  $\tilde{\chi}_2^0$  events

## > LOI sample:

- Signal generated with `Whizard1.51`  
Background generated with `Whizard1.40`
- The RDR beam spectrum was used

## > DBD sample:

- Signal (as well as SM background) generated with `Whizard 1.95`
- The TDR beam spectrum was used

▪ **Note:** in the signal samples, the  $M_W$  was inadvertently lowered by Whizard to  $M_W = 79.8$  GeV

- Signal + background were simulated and reconstructed with `ilcsoft v01-06`
- The jet energy scale was increased by 1%
- No  $\gamma\gamma$  background overlay
- The analysis was re-run on existing data samples

- Signal + background were simulated and reconstructed with `ilcsoft v01-16-02`
- The jet energy scale was **not** increased
- The  **$\gamma\gamma$  background overlay** was taken into account
- The analysis was re-run



# Analysis Strategy

- Remove  $\gamma\gamma \rightarrow \text{hadrons}$  background
  - Cluster event into 4 jets (Durham)
  - Run kinematic fit (equal mass constraint)
  - Run isolated lepton finder (Junping)
  - Perform SUSY preselection
  - Separate  $\tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  samples
  - Perform mass measurement
- } **only for DBD**
- } **only for DBD**
- } **common to both LOI and DBD**



# Analysis Strategy

- Remove  $\gamma\gamma \rightarrow \text{hadrons}$  background
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- only for DBD**
- only for DBD**
- common to both LOI and DBD**



# Removing the $\gamma\gamma$ Background II

- Tested configurations:

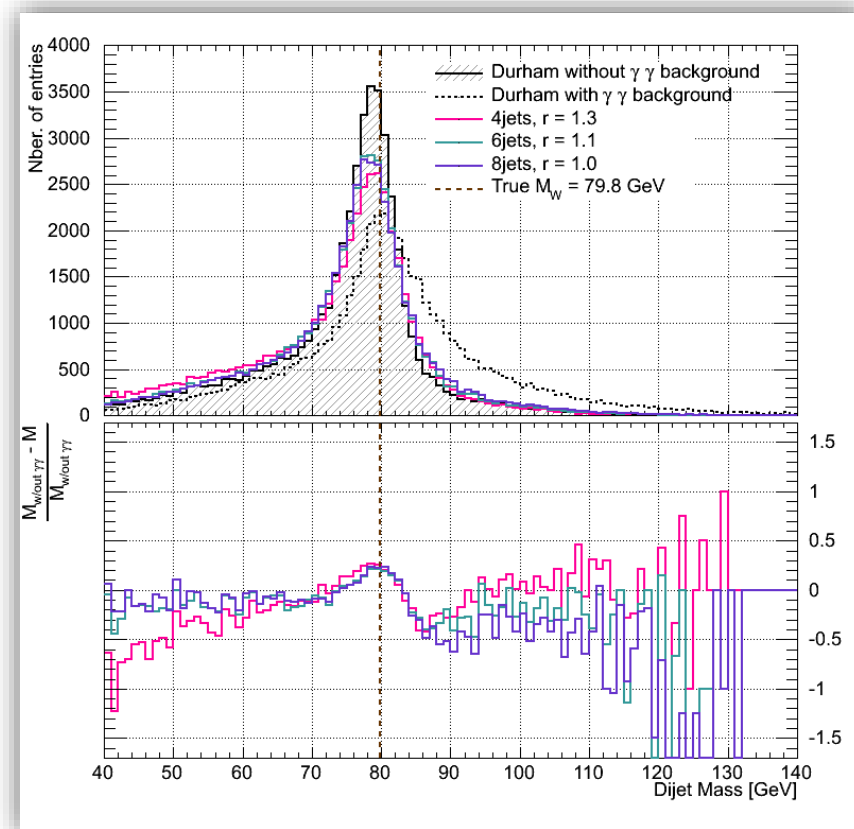
| Jets #  | 4 jets     | 6 jets     | 8 jets     |
|---------|------------|------------|------------|
|         | 1.1        | 0.9        | 0.8        |
| R value | <b>1.3</b> | <b>1.1</b> | <b>1.0</b> |
|         | 1.5        | 1.3        | 1.2        |

- Find the configuration most similar to the Durham distribution  $\rightarrow \chi^2$  test :

| Jets #        | R val.     | $\chi^2/\text{ndf}$<br>W | $\chi^2/\text{ndf}$<br>Z |
|---------------|------------|--------------------------|--------------------------|
| 4 jets        | 1.3        | 13.4                     | 11.6                     |
| <b>6 jets</b> | <b>1.1</b> | <b>6.9</b>               | <b>4.7</b>               |
| 8 jets        | 1.0        | 9.3                      | 6.8                      |

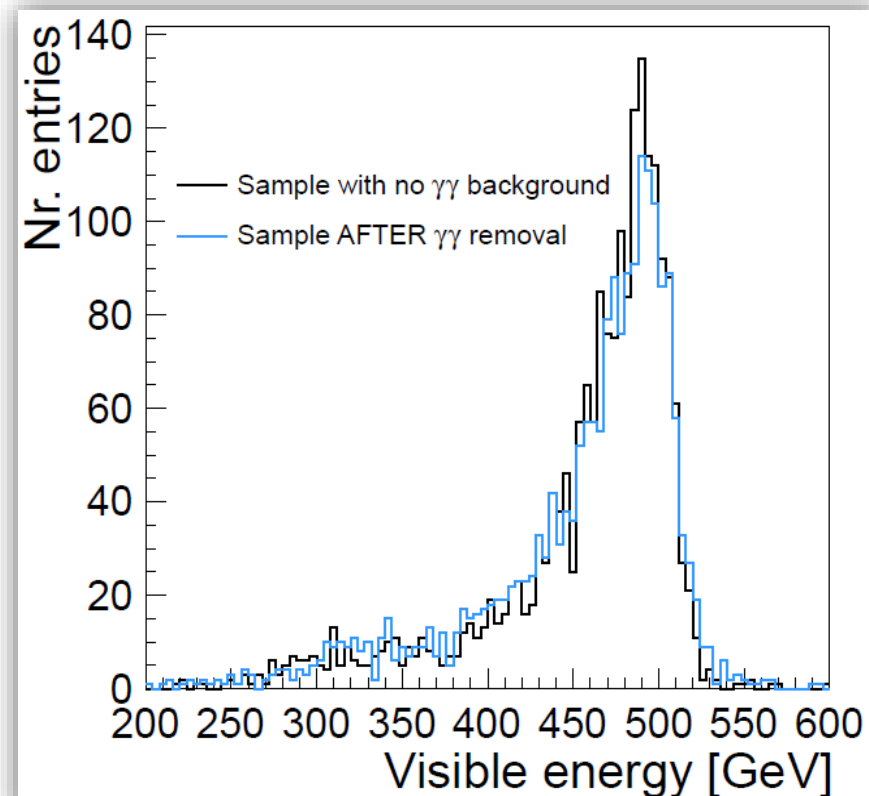
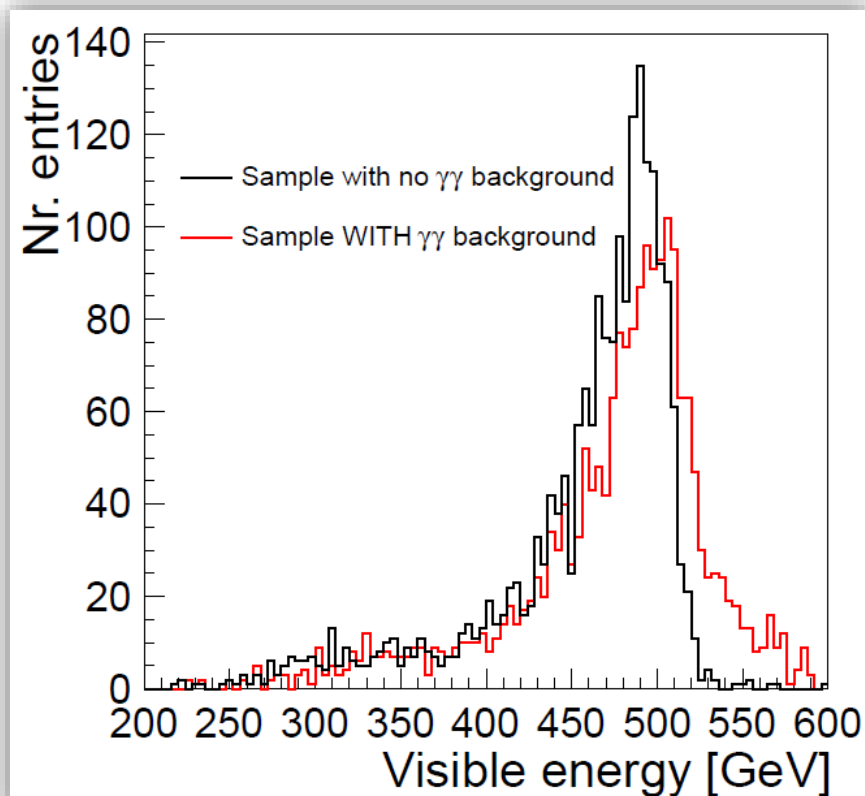
- The 6 jets configuration with an R-value of 1.1 is best for  $\gamma\gamma$  background removal !

Used: **exclusive longitudinal  $k_T$  algorithm**



# Removing the $\gamma\gamma$ Background III

- Checking the  $\gamma\gamma$  background removal for the chosen configuration:



- After running the longitudinal exclusive  $k_T$  algorithm the visible energy is very similar to the no background case.





# Analysis Strategy

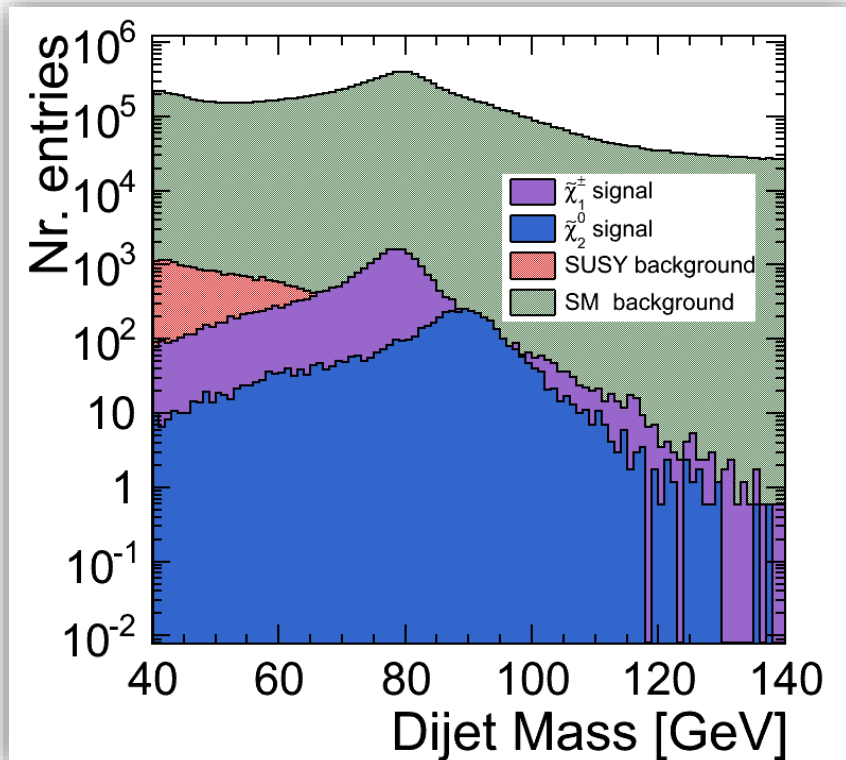
- Remove  $\gamma\gamma \rightarrow \text{hadrons}$  background
  - Cluster event into 4 jets (Durham)
  - Run kinematic fit (equal mass constraint)
  - Run isolated lepton finder (Junping)
  - Perform SUSY preselection
  - Separate  $\tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  samples
  - Perform mass measurement
- only for DBD
- only for DBD
- common to both LOI and DBD



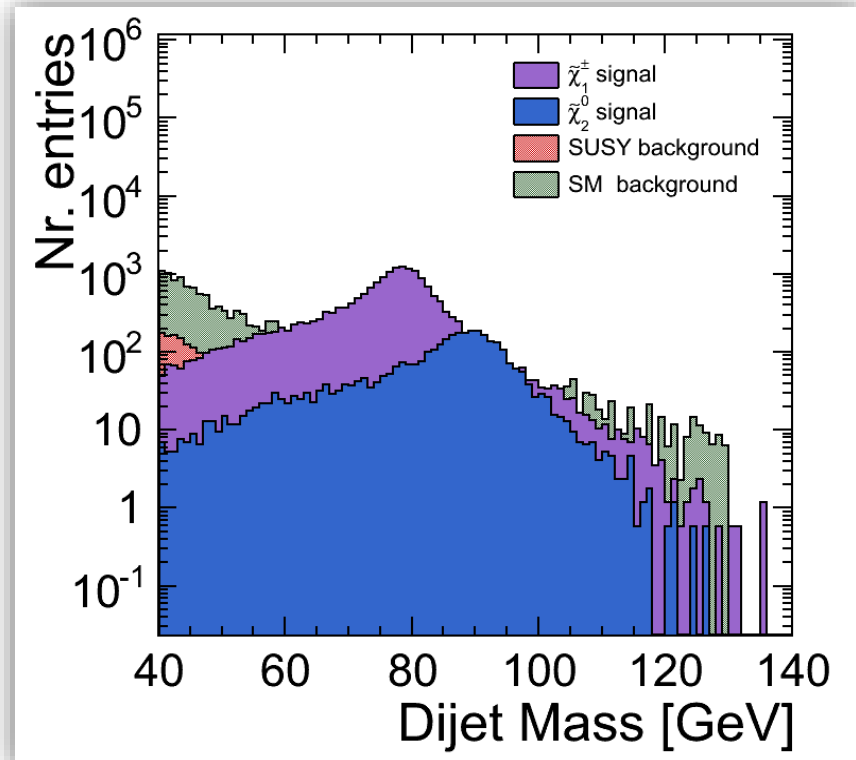
# Preselection Outcome

Example: the DBD sample [ LOI sample very similar]

Before preselection



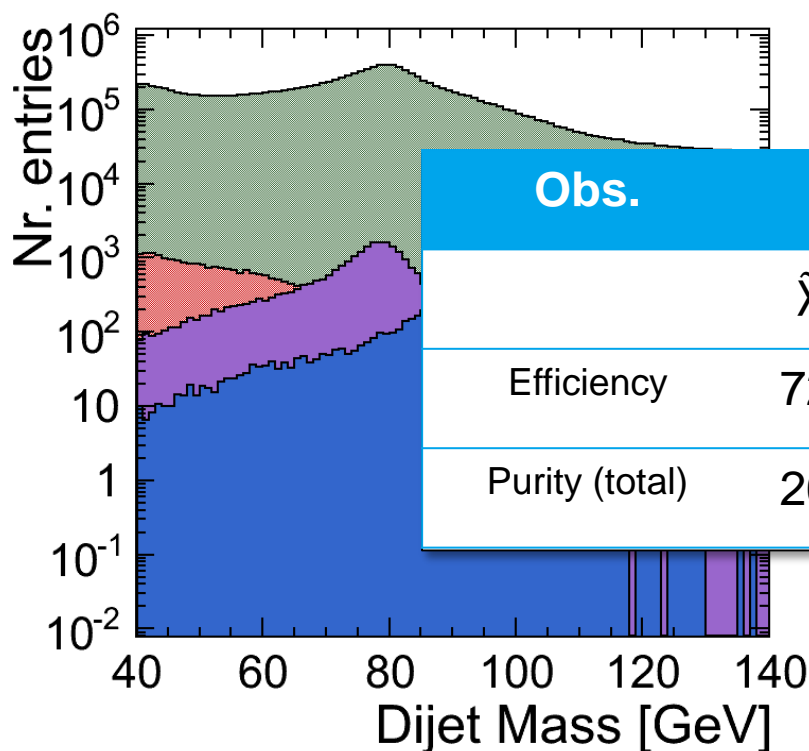
After preselection



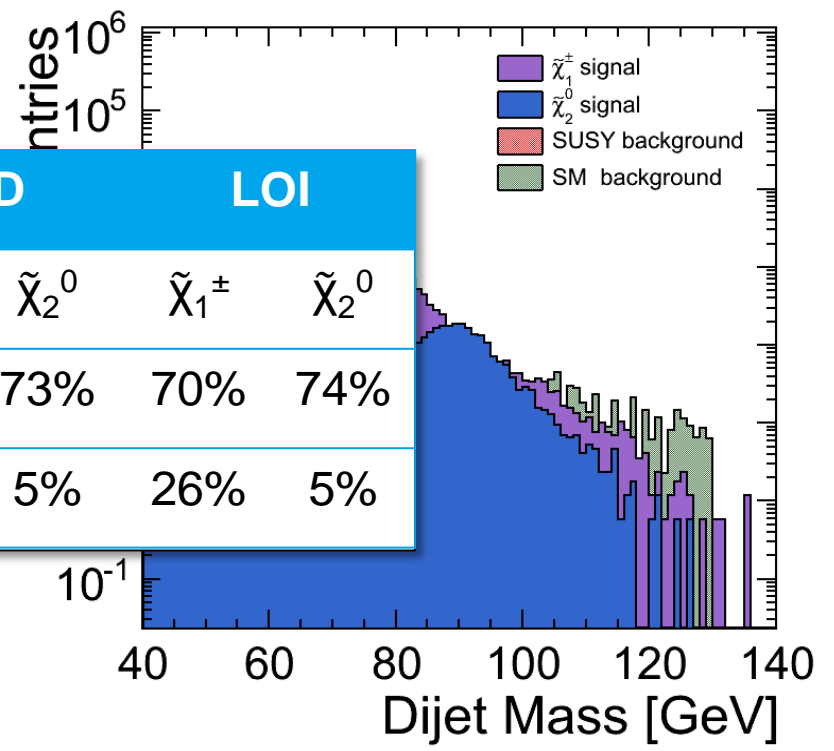
# Preselection Outcome

Example: the DBD sample [ LOI sample very similar]

Before preselection



After preselection



|                | Obs. | DBD                  |                    | LOI                  |                    |
|----------------|------|----------------------|--------------------|----------------------|--------------------|
|                |      | $\tilde{\chi}_1^\pm$ | $\tilde{\chi}_2^0$ | $\tilde{\chi}_1^\pm$ | $\tilde{\chi}_2^0$ |
| Efficiency     |      | 72%                  | 73%                | 70%                  | 74%                |
| Purity (total) |      | 26%                  | 5%                 | 26%                  | 5%                 |



# Analysis Strategy

- Remove  $\gamma\gamma \rightarrow \text{hadrons}$  background
  - Cluster event into 4 jets (Durham)
  - Run kinematic fit (equal mass constraint)
  - Run isolated lepton finder (Junping)
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  - Separate  $\tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  samples
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- } only for DBD
- } only for DBD
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# $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_2^0$ Signal Sample Further Separation

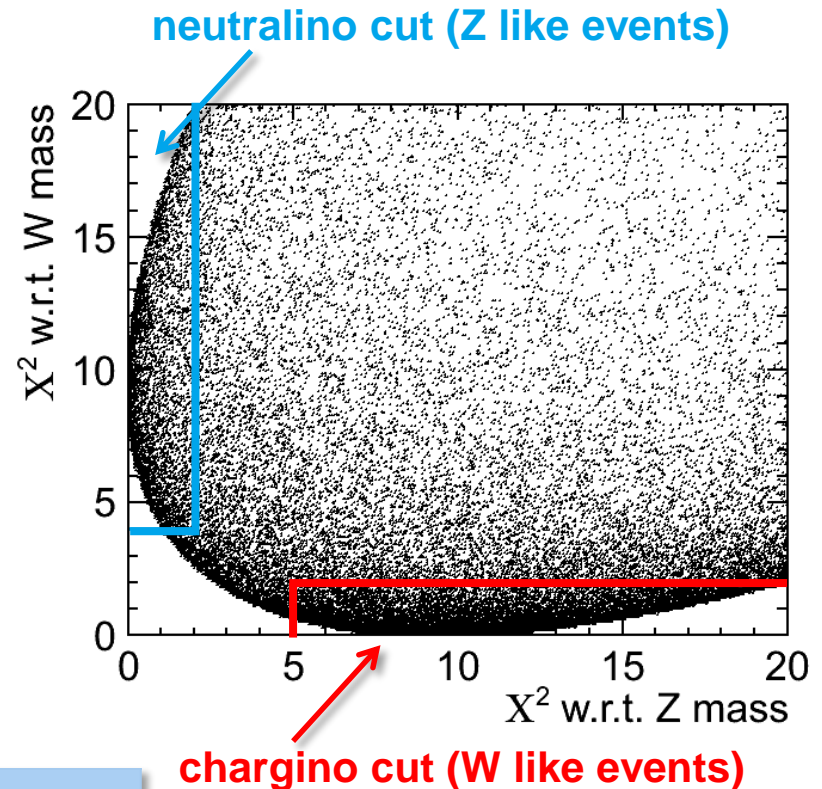
- > Calculate  $\chi^2$  with respect to nominal W / Z mass

$$\chi^2(m_{j1}, m_{j2}) = \frac{(m_{j1} - m_V)^2 + (m_{j2} - m_V)^2}{\sigma^2}$$



$\min \chi^2 \rightarrow \tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  separation

- > Downside: lose statistics
  - Cut away 43% of  $\tilde{\chi}_1^\pm$  surviving events
  - Cut away 68% of  $\tilde{\chi}_2^0$  surviving events
- > However, after the  $\chi^2$  cut, the separation is quite clear:



| Obs.           | DBD                  |                    | LOI                  |                    |
|----------------|----------------------|--------------------|----------------------|--------------------|
|                | $\tilde{\chi}_1^\pm$ | $\tilde{\chi}_2^0$ | $\tilde{\chi}_1^\pm$ | $\tilde{\chi}_2^0$ |
| Efficiency     | 53%                  | 30%                | 56%                  | 34%                |
| Purity (total) | 63%                  | 38%                | 62%                  | 35%                |
| Purity (SUSY)  | 94%                  | 62%                | 95%                  | 66%                |

# $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_2^0$ Mass Measurement

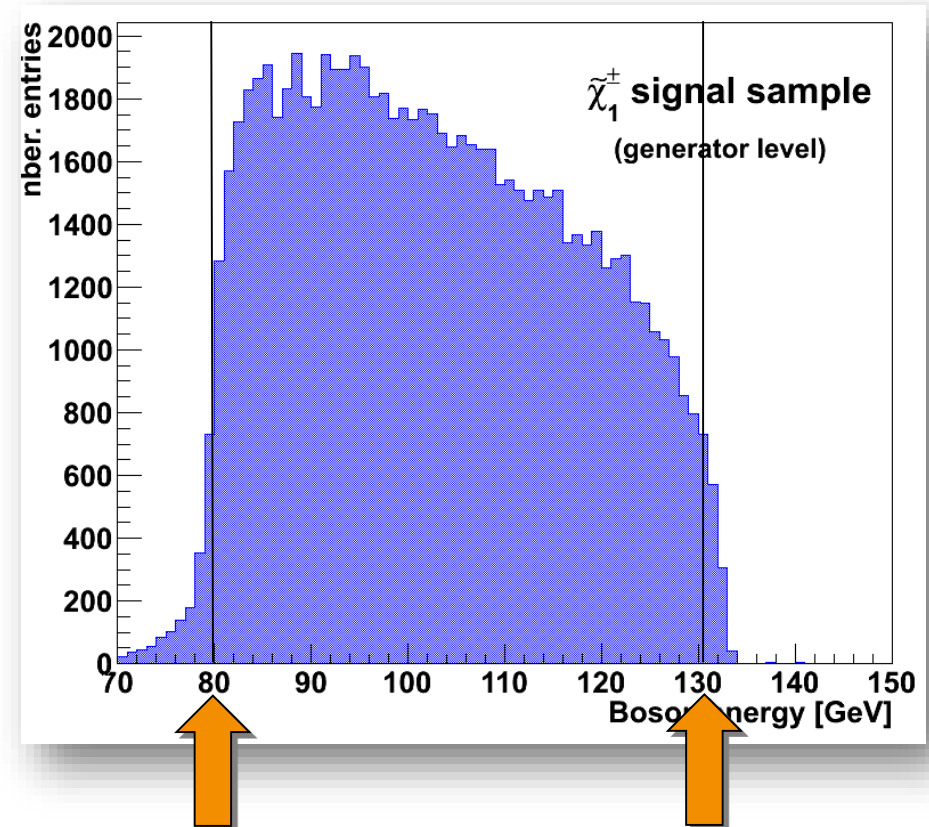
- > Mass difference to LSP ( $\tilde{\chi}_1^0$ ) is **larger** than  $M_Z$
- > Observe the decays of real gauge bosons
- > 2 body decay  $\rightarrow$  the edges of the energy spectrum are kinematically determined
- > **Use dijet energy spectrum „end points“ in order to calculate masses**

$$\gamma = \frac{E_{beam}}{M_\chi}$$

$$E_\pm = \gamma \cdot E_V^* \pm \gamma \cdot \beta \cdot \sqrt{E_V^{*2} - M_V^2}$$

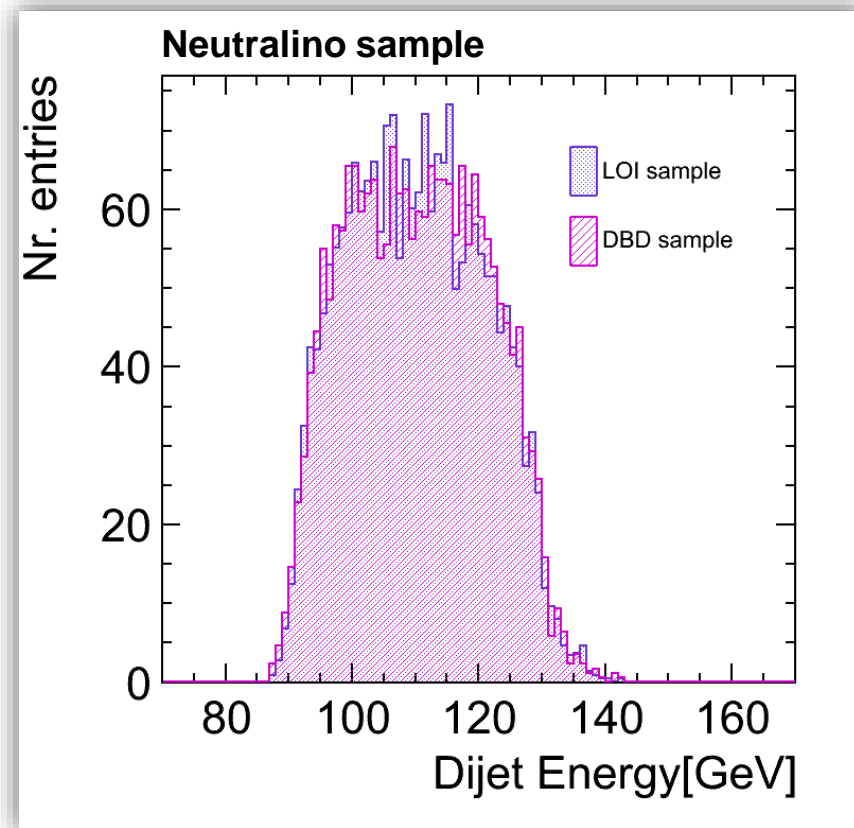
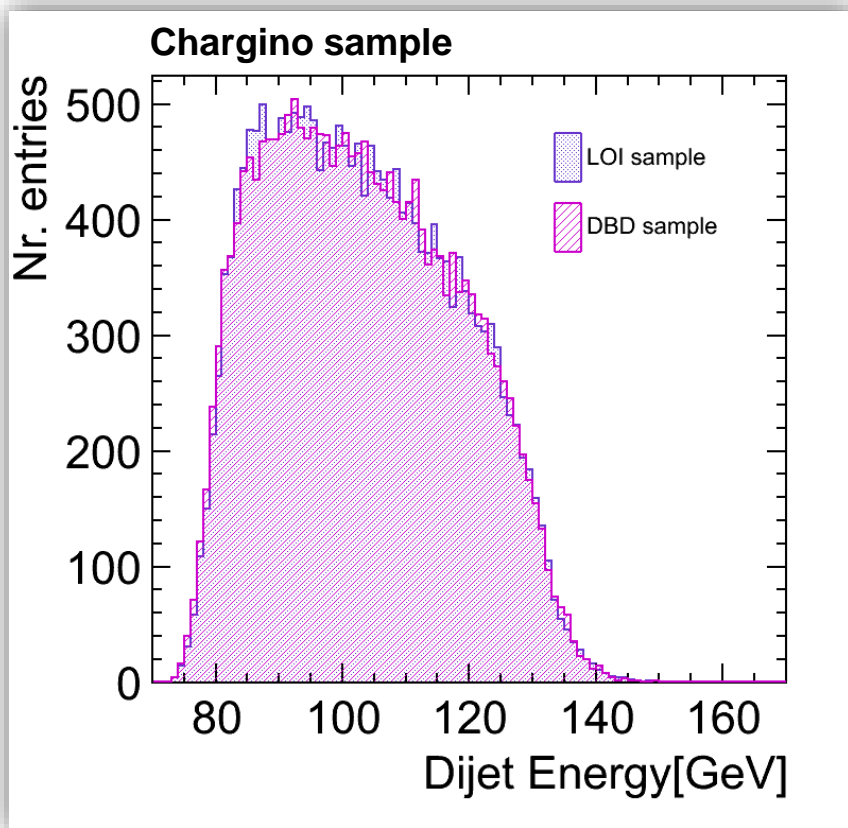
Real edge values [GeV]:

| $W_{low}$ | $W_{high}$ | $Z_{low}$ | $Z_{high}$ |
|-----------|------------|-----------|------------|
| 80.17     | 131.53     | 93.24     | 129.06     |



# Dijet [Boson] Energy Comparison LOI - DBD

➤ Use dijet energy to measure  $\tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  mass

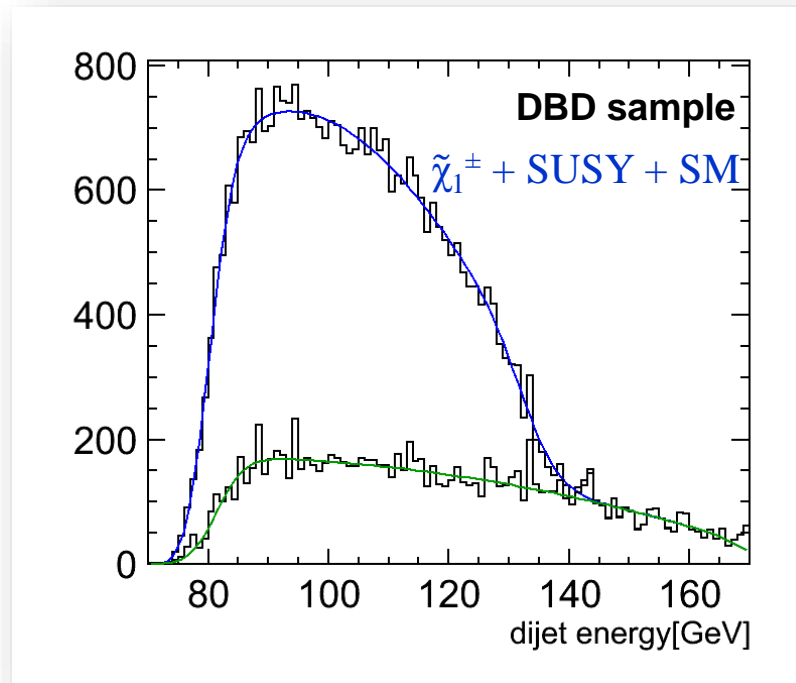


➤ The DBD distribution appears slightly shifted towards lower energies. Nevertheless, **the two distributions agree very well.**

# $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_2^0$ Mass Measurement – “Endpoint” Method

- Fit dijet energy spectrum and obtain edge positions:

$$f(x; t_0, b_0, \sigma_1, \gamma) = f_{SM} + \int_{t_0}^{t_1} (b_2 t^2 + b_1 t + b_0) V(x - t, \sigma(t), \gamma) dt$$

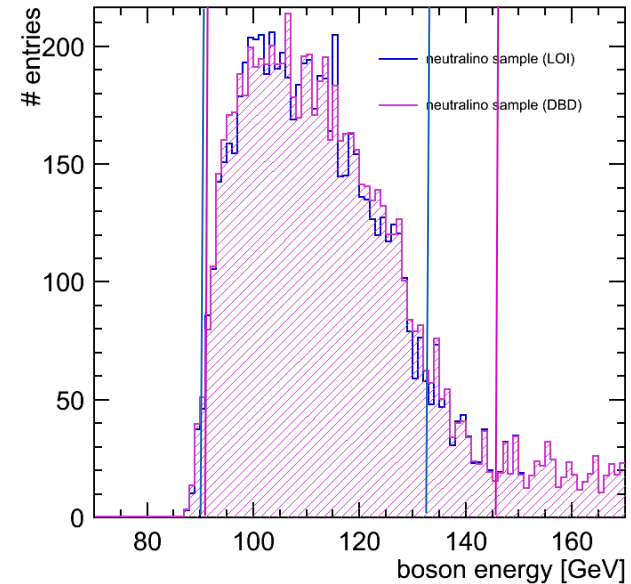
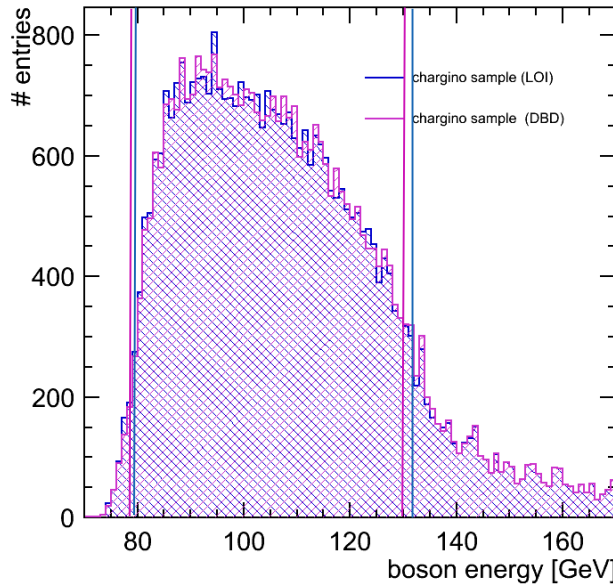


Where:

- The polynomial accounts for the slope of the initial spectrum
- The Voigt function accounts for the detector resolution and gauge boson width



# Issues of the “Endpoint Method”



| Sim. | Edge $W_{\text{low}}$ [GeV] | Edge $W_{\text{high}}$ [GeV] | Edge $Z_{\text{low}}$ [GeV] | Edge $Z_{\text{high}}$ [GeV] |
|------|-----------------------------|------------------------------|-----------------------------|------------------------------|
| DBD  | $79.5 \pm 0.5$              | $130.2 \pm 1.1$              | $91.3 \pm 0.6$              | $146.1 \pm 4.8$              |
| LOI  | $79.7 \pm 0.3$              | $131.9 \pm 0.9$              | $91.0 \pm 0.7$              | $133.6 \pm 0.5$              |

The fitting method appears to be highly dependent on small changes in the fitted distribution → it is **NOT** appropriate for comparing the two samples.

**We need to apply a different edge extraction method!**

# Endpoint Extraction using an FIR Filter

- Finite Impulse Response (FIR) filters are digital filters used in signal processing.
- FIR filters can operate both on discrete as well as continuous values.
- The concept of “finite impulse response” ↔ **the filter output** is computed as a finite, weighted sum of a finite number of values from the filter input.

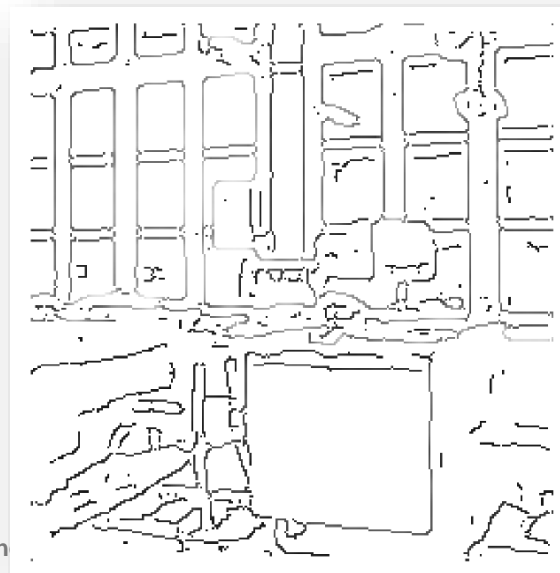
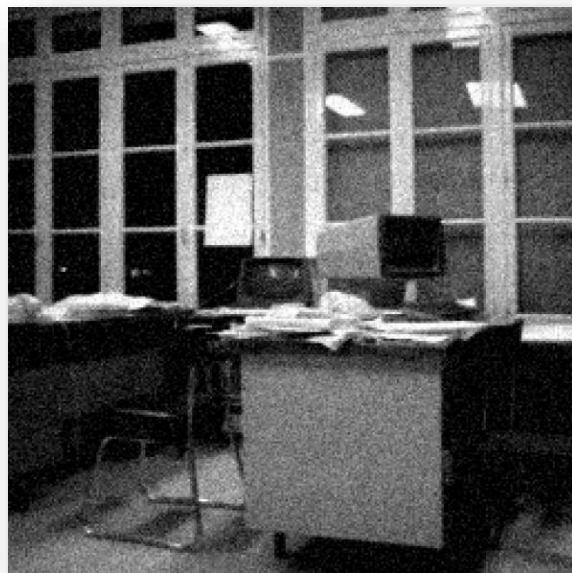
$$y[n] = \sum_{k=-M_1}^{M_2} b_k x[n-k]$$

the input signal

the filter coefficients (weights)

- $y$  is obtained by convolving the input signal with the (finite) weights
- FIR filters are used to detect edges in image processing techniques:

D. Demigny, T. Kamlé

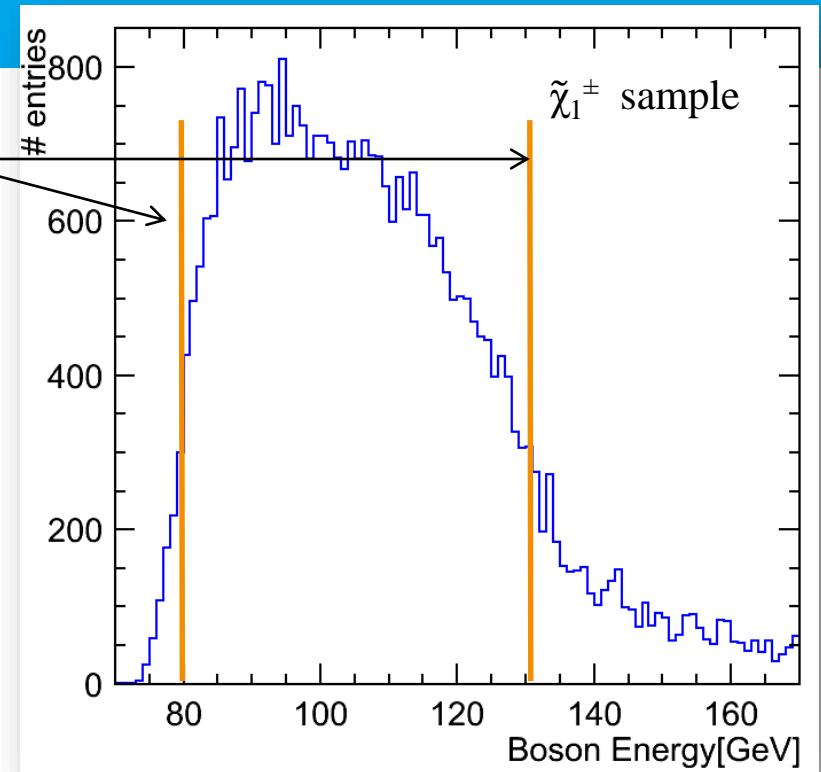


Madalina Ch



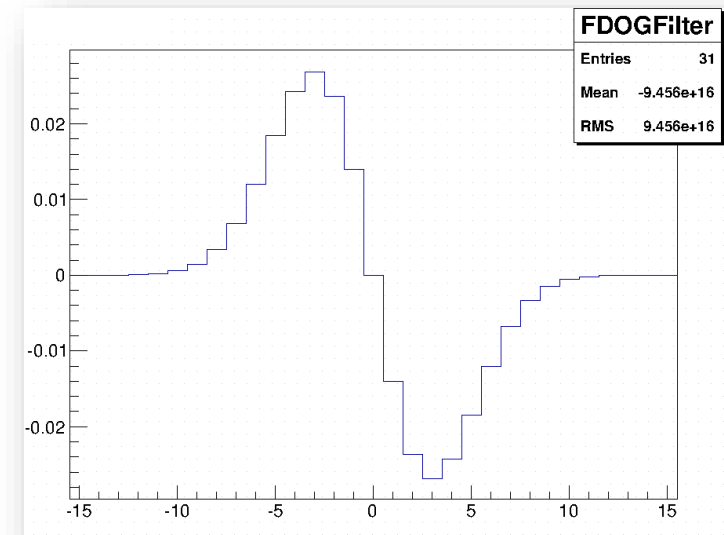
# Applying an FIR Filter

- > Goal: find edge positions in spectrum



# Applying an FIR Filter

- > Goal: find edge positions in spectrum
- > Strategy:
  - Choose an FIR filter
  - Note: **filter length** << **signal histogram length**

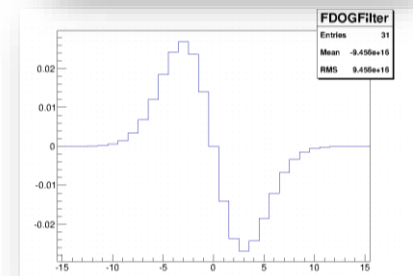
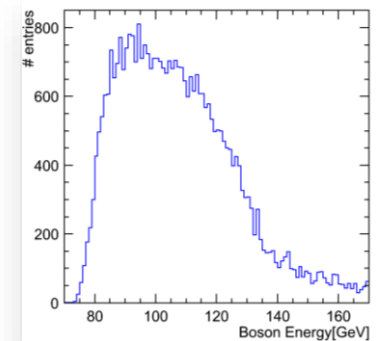


# Applying an FIR Filter

- > Goal: find edge positions in spectrum
- > Strategy:
  - Choose an FIR filter
  - Note: filter length  $\ll$  signal histogram length
  - Treat both signal histogram as well as filter as **arrays**:

|        |   |    |    |     |    |    |     |
|--------|---|----|----|-----|----|----|-----|
| Bin #  | 1 | 2  | 3  | ... | 98 | 99 | 100 |
| Signal | 0 | 15 | 28 | ... | 34 | 22 | 4   |

|        |   |      |      |     |       |       |    |
|--------|---|------|------|-----|-------|-------|----|
| Bin #  | 1 | 2    | 3    | ... | 28    | 29    | 30 |
| Filter | 0 | 0.01 | 0.02 | ... | -0.02 | -0.01 | 0  |



# Applying an FIR Filter

> Goal: find edge positions in spectrum

> Strategy:

- Choose an FIR filter
- Note: filter length  $\ll$  signal histogram length
- Treat both signal histogram as well as filter as **arrays**
- Calculate **dot product between Signal and Filter** → obtain one value

|        |   |    |    |     |    |    |     |
|--------|---|----|----|-----|----|----|-----|
| Bin #  | 1 | 2  | 3  | ... | 98 | 99 | 100 |
| Signal | 0 | 15 | 28 | ... | 34 | 22 | 4   |

|        |   |      |      |     |       |       |    |
|--------|---|------|------|-----|-------|-------|----|
| Bin #  | 1 | 2    | 3    | ... | 28    | 29    | 30 |
| Filter | 0 | 0.01 | 0.02 | ... | -0.02 | -0.01 | 0  |

$$0 \times 0 + 0.01 \times 15 + 0.02 \times 28 + \dots = \text{val1}$$



# Applying an FIR Filter

> Goal: find edge positions in spectrum

> Strategy:

- Choose an FIR filter
- Note: filter length  $\ll$  signal histogram length
- Treat both signal histogram as well as filter as arrays
- Calculate dot product between Signal and Filter  $\rightarrow$  obtain one value

|        |   |    |    |     |    |    |     |
|--------|---|----|----|-----|----|----|-----|
| Bin #  | 1 | 2  | 3  | ... | 98 | 99 | 100 |
| Signal | 0 | 15 | 28 | ... | 34 | 22 | 4   |

|        |   |      |      |     |       |       |    |
|--------|---|------|------|-----|-------|-------|----|
| Bin #  | 1 | 2    | 3    | ... | 28    | 29    | 30 |
| Filter | 0 | 0.01 | 0.02 | ... | -0.02 | -0.01 | 0  |



$$0 \times 15 + 0.01 \times 28 + \dots = \text{val2}$$

- **“Move”** Filter along the (length) of the signal  $\rightarrow$  obtain more values, which will form the total filter response

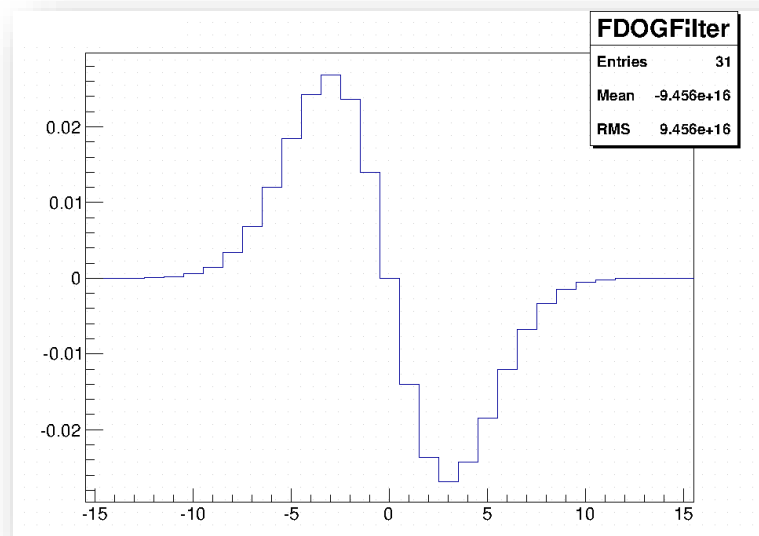
# Choosing the Appropriate Filter

> In order to choose an appropriate filter one can apply the following criteria:

Canny's criteria: [J. F. Canny. **A computational approach to edge detection.** *IEEE Trans. Pattern Analysis and Machine Intelligence*, pages 679-698, 1986]

- Good detection: probability of obtaining a peak in the response must be high
- Localisation: standard deviation of the peak position must be small
- Multiple response minimisation: probability of false positive detection must be small

> Canny has suggested that an optimal filter is very similar to the  
first derivative of a Gaussian





# Optimising the FDOG Filter

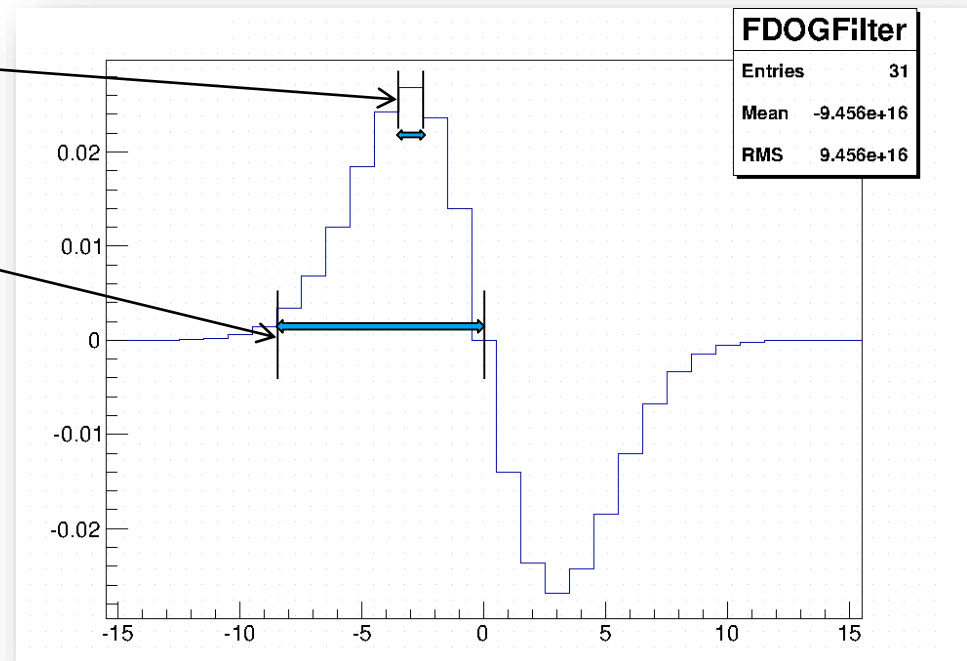
➤ There are two important filter characteristics that must be optimised:

the bin size

the filter size

It is crucial to strike the right balance between the two:

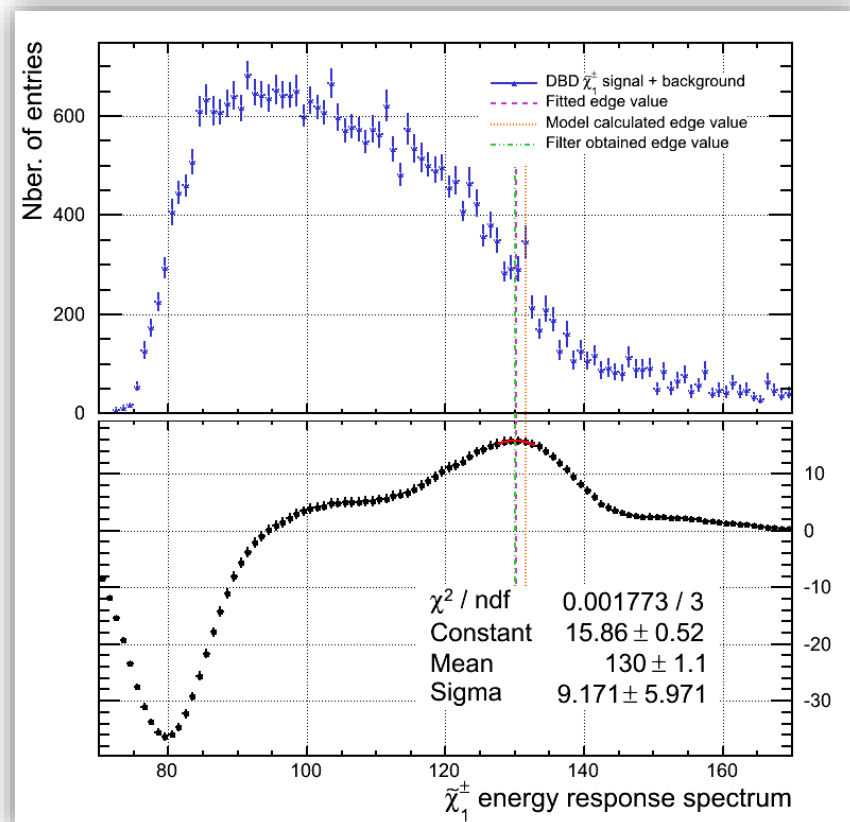
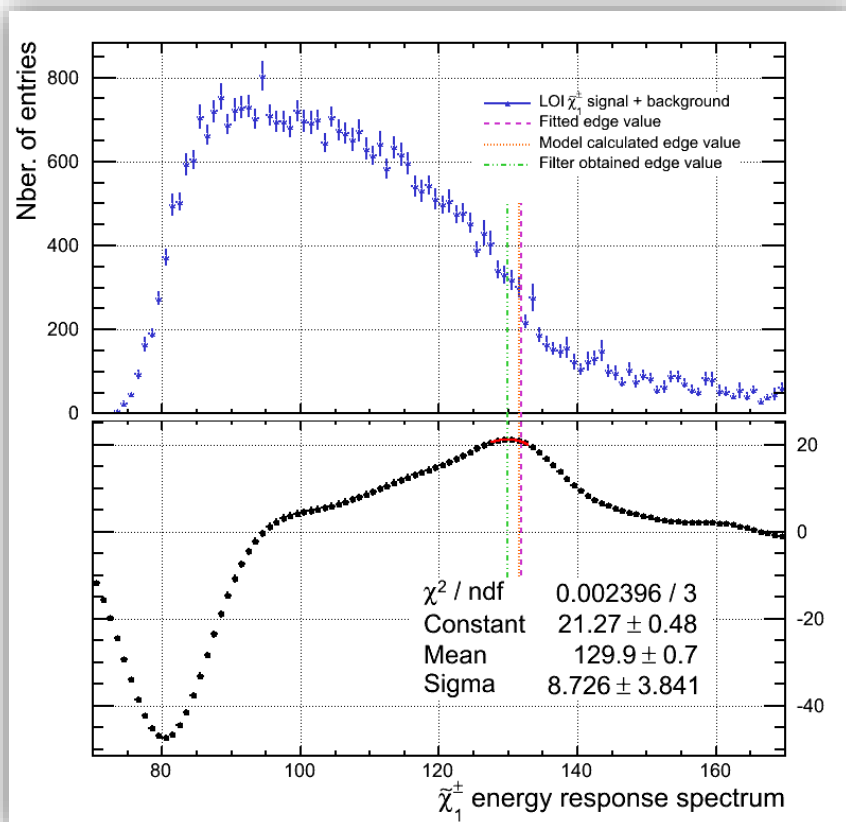
- If the bin size is too small → the filter picks up a lot of statistical fluctuations
- If the filter size is too large → the edge position cannot be localised anymore



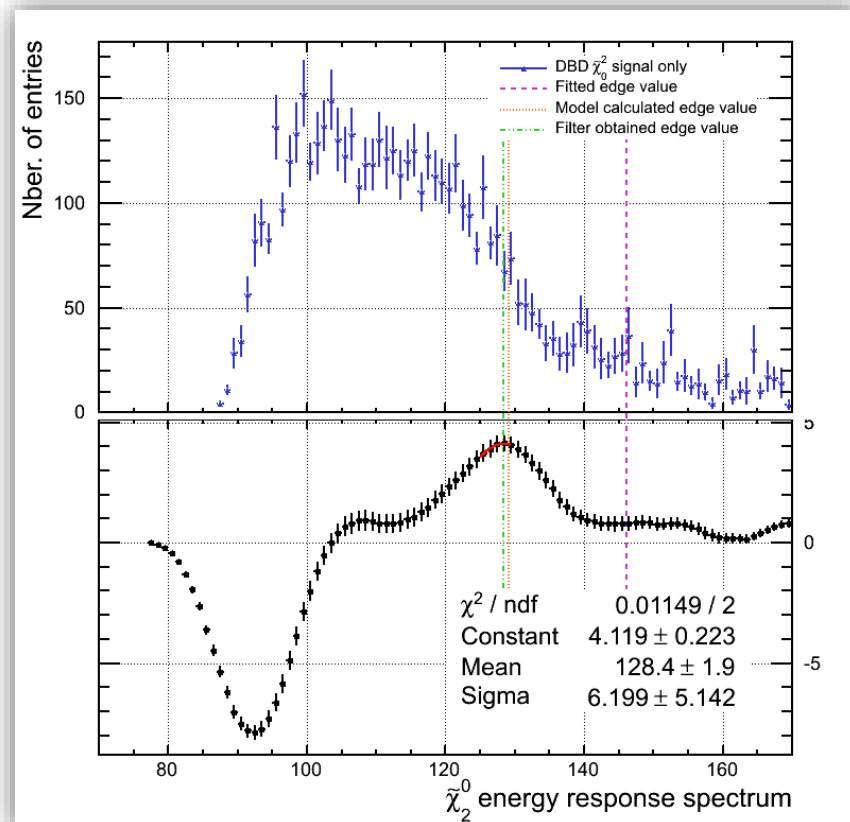
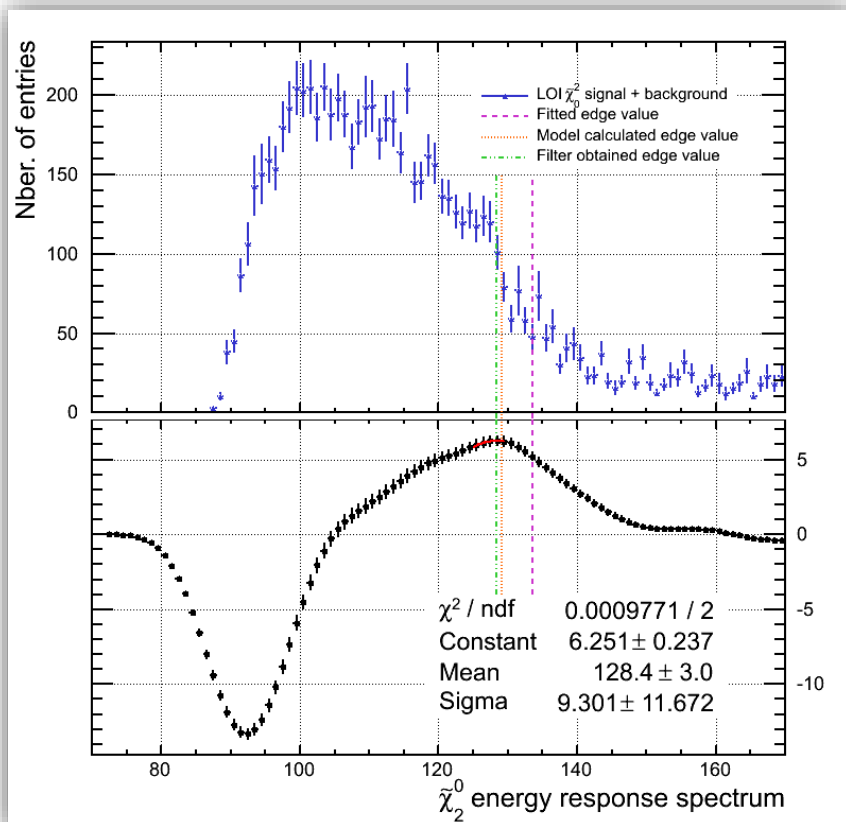
A toy MC study has been performed to optimise the filter and bin size.

Chosen values: bin size = 1 GeV/Bin; filter size = 5 x 2 bins.

# FIR Edge Extraction Comparison – LOI to DBD

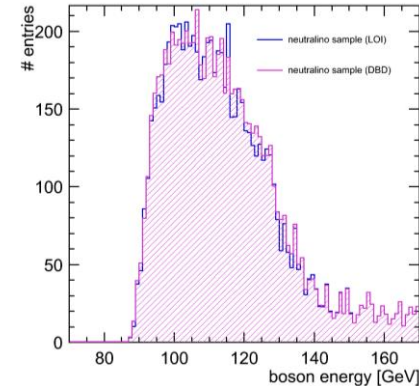
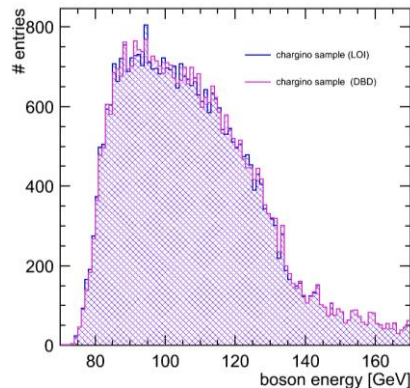


# FIR Edge Extraction Comparison – LOI to DBD



- On the chargino samples the filter appears to perform slightly worse than the fit in the LOI sample and just as well in the DBD case.
- However the filter performs considerably better in the case of the neutralino samples both for the LOI and the DBD case.

# Edge Extraction Comparison



| True   |     | 80.17                       | 131.53                       | 93.24                       | 129.06                       |
|--------|-----|-----------------------------|------------------------------|-----------------------------|------------------------------|
| Sim.   |     | Edge $W_{\text{low}}$ [GeV] | Edge $W_{\text{high}}$ [GeV] | Edge $Z_{\text{low}}$ [GeV] | Edge $Z_{\text{high}}$ [GeV] |
| filter | LOI | $79.7 \pm 0.3$              | $131.9 \pm 0.9$              | $91.0 \pm 0.7$              | $133.6 \pm 0.5$              |
|        | DBD | $79.5 \pm 0.5$              | $130.2 \pm 1.1$              | $91.3 \pm 0.6$              | $146.1 \pm 4.8$              |
|        | LOI | $80.4 \pm 0.2$              | $129.9 \pm 0.7$              | $92.3 \pm 0.4$              | $128.3 \pm 0.9$              |
|        | DBD | $79.8 \pm 0.3$              | $129.9 \pm 1.0$              | $92.2 \pm 0.4$              | $128.3 \pm 0.6$              |

The filter extraction method is preferable:

- it is more stable
- provides smaller uncertainties in determining the edge position.

# Edge Extraction Comparison

filter

| True | 80.17                       | 131.53                       | 93.24                       | 129.06                       |
|------|-----------------------------|------------------------------|-----------------------------|------------------------------|
| Sim. | Edge $W_{\text{low}}$ [GeV] | Edge $W_{\text{high}}$ [GeV] | Edge $Z_{\text{low}}$ [GeV] | Edge $Z_{\text{high}}$ [GeV] |
| LOI  | 80.4±0.2                    | 129.9±0.7                    | 92.3±0.4                    | 128.3±0.9                    |
| DBD  | 79.8±0.3                    | 129.9±1.0                    | 92.2±0.4                    | 128.3±0.6                    |

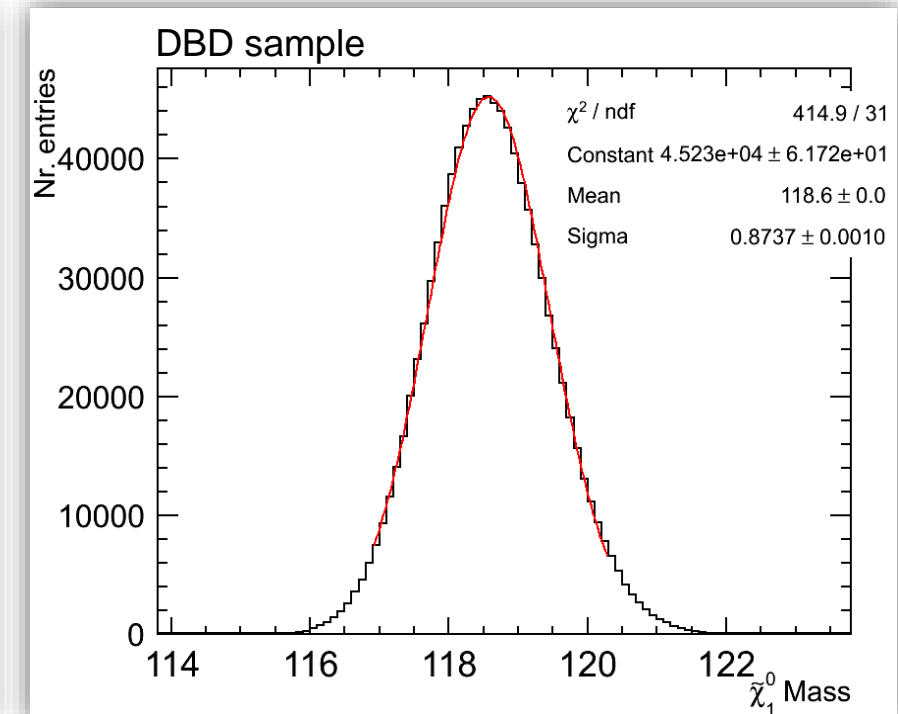
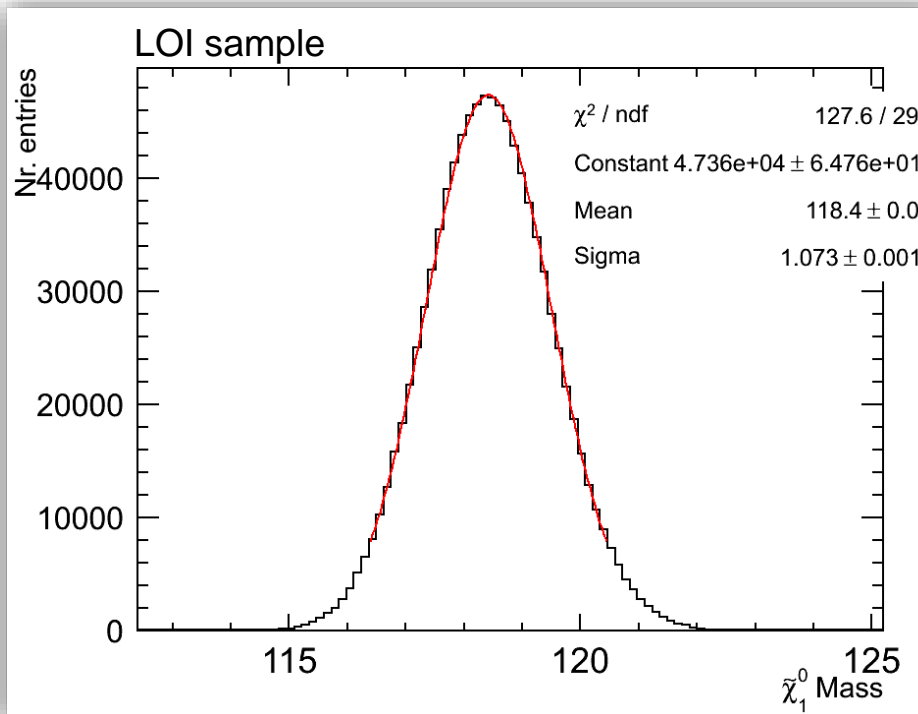
| Sample      | Mass $\tilde{\chi}_1^\pm$ [GeV] | Mass $\tilde{\chi}_2^0$ [GeV] | Mass $\tilde{\chi}_1^0$ [GeV] |
|-------------|---------------------------------|-------------------------------|-------------------------------|
| <b>TRUE</b> | <b>216.5</b>                    | <b>216.7</b>                  | <b>115.7</b>                  |
| LOI         | 216.9±3.2                       | 220.0±1.4                     | 118.4±1.1                     |
| DBD         | 217.3±3.2                       | 220.4±1.5                     | 118.5±0.9                     |

- The filter method is much more stable in determining the edge position
- The mass values extracted from the LOI and DBD samples are compatible within their statistical errors
- The systematic errors will be addressed by a mass calibration study [ongoing]



# Toy MC for the Mass Calculation

- To estimate the statistical precision of the mass measurement → toy MC
- Input: edge values + their fluctuations as obtained from the filter
- 1 000 000 values were generated within the edge fluctuations
- The mass calculations have been performed with the generated values 1 000 000 times



# Conclusions

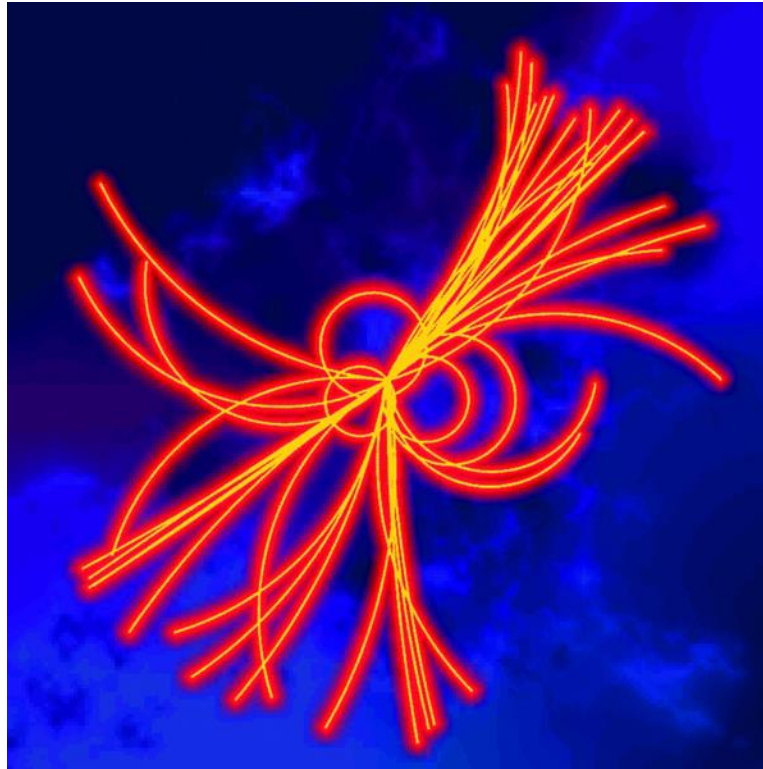
- A comparison between the LOI and DBD detectors has been made with “Point5”:
  - The fitting method for the mass determination appears very sensitive to small changes. A more robust method is needed.
  - Applying a finite impulse response (FIR) filter in order to extract the edge information instead of the fitting method is:
    - More robust (i.e. independent on distribution shape)
    - Provides just as good if not better statistical precision
  - The  $\gamma\gamma$  background in the DBD sample successfully removed

## > Outlook:

- Perform mass calibration (to determine systematics).
- Perform 2D fit on dijet masses to improve the x-section measurement
- Perform full comparison LOI – DBD – SGV.



# Thank You!





# Preselection

> Apply the following cuts to both samples:

1. Number tracks in event  $> 20$
2.  $100 \text{ GeV} < E_{\text{visible}} < 300 \text{ GeV}$
3.  $E_{\text{jet}} > 5 \text{ GeV}$
4.  $|\cos(\theta_{\text{jets}})| < 0.9$
5.  $Y_{34} > 0.001$
6. Number tracks per jet  $> 2$
7.  $|\cos(\theta_{\text{miss}})| < 0.99$
8.  $E_{\text{lepton}} < 25 \text{ GeV}$
9. Number of PFOs per jet  $> 3$
10.  $|\cos(\theta_{\text{miss}})| < 0.8$
11.  $M_{\text{miss}} > 220 \text{ GeV}$
12. Kinematic fit converged

13. No isolated lepton
14.  $30 < \text{Number PFOs in event} < 150$
15.  $4 < \text{Nr. Tracks with } P_T > 1 \text{ GeV} < 50$
16. Thrust  $< 0.98$

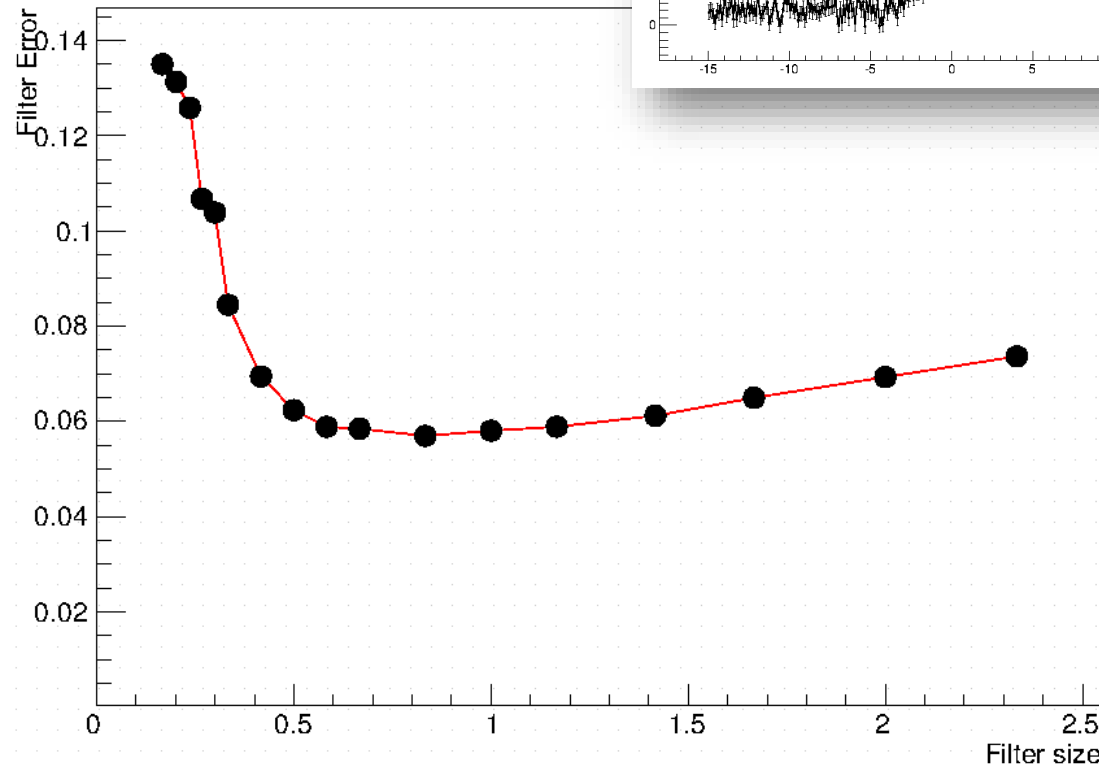
DBD

LOI & DBD common



# Optimising the FDOG Filter

Studied the effect of the filter size on a smeared step edge Monte Carlo data.

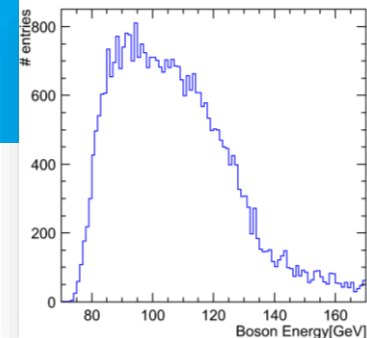
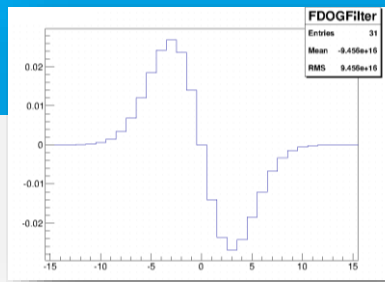


S. Caiazza

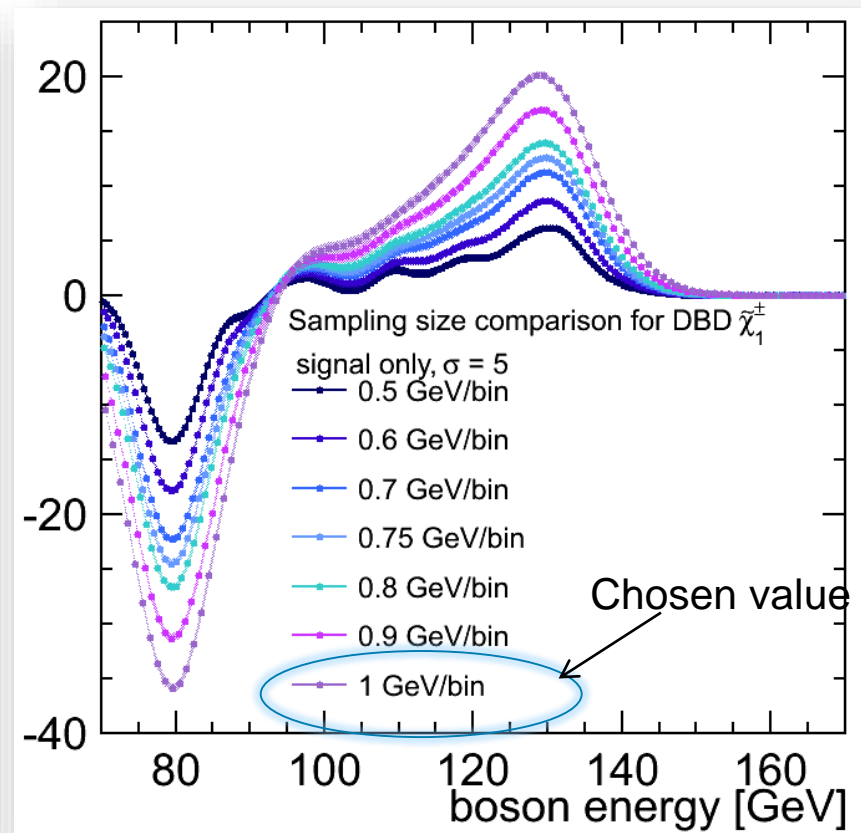
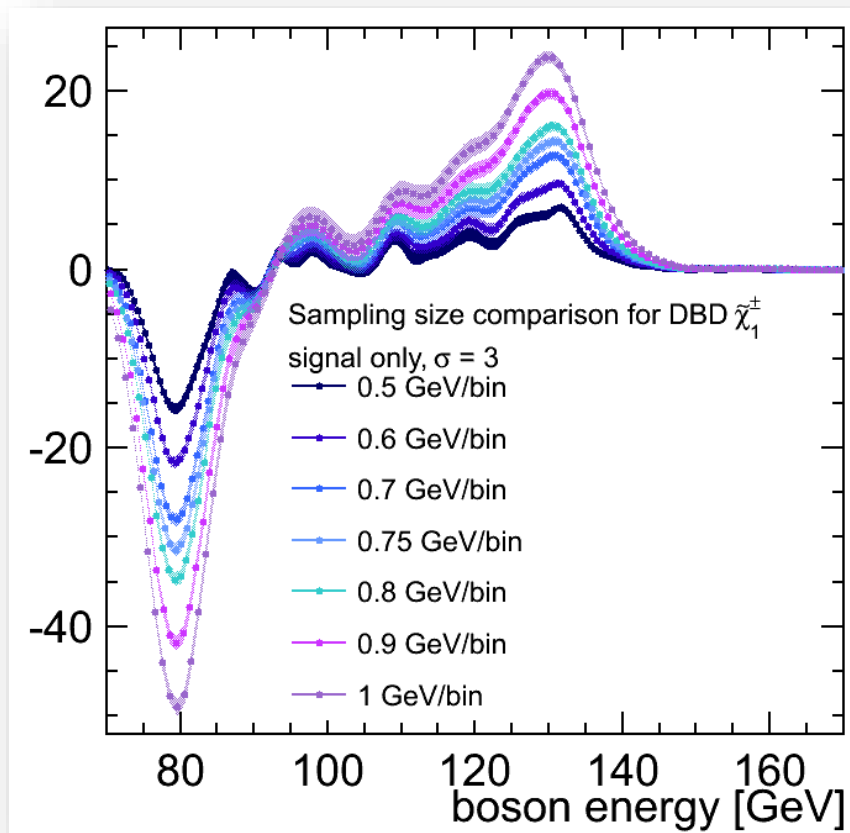
The FDOG filter does indeed perform best.  
The filter size should be comparable to the size of the edge feature.  
We chose  $\sigma = 5$  bins.

# Optimising the FDOG Filter

- > There are two important filter characteristics that must be optimised: the **bin size** and the filter size.



Filter response after applying the FDOG Filter to the  $\tilde{\chi}_1^\pm$  energy distribution:



# Back up slides

# Removing the $\gamma\gamma$ Background

➤ Use the **longitudinal exclusive  $k_T$**  jet clustering algorithm:

➤ It calculates:

➤ The “distance” between each pair of reconstructed particles:

$$dist_{ij} = \frac{\min(p_{Ti}^2, p_{Tj}^2) \cdot \Delta \mathbf{R}_{ij}^2}{R^2}$$

➤ The distance between each reconstructed particle and the beam ( $dist_{ij}^{Beam}$ )

➤ If the  $dist_{ij}^{Beam}$  is minimum then the particle is discarded

➤ The number of required jets as well as the R parameter are free parameters.

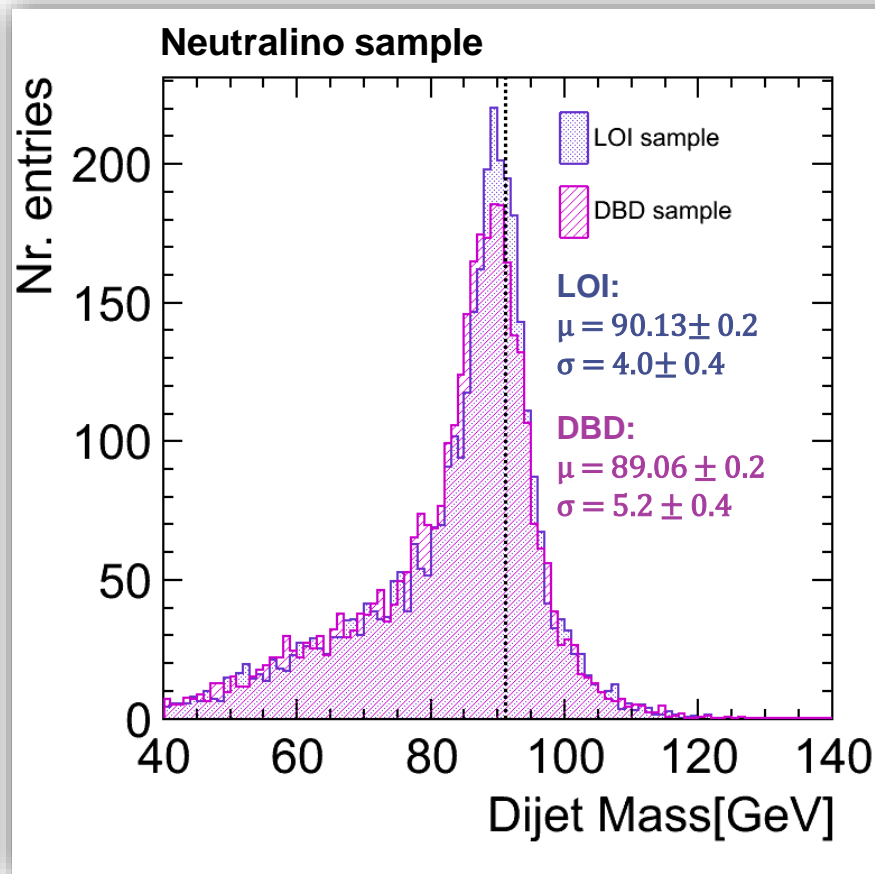
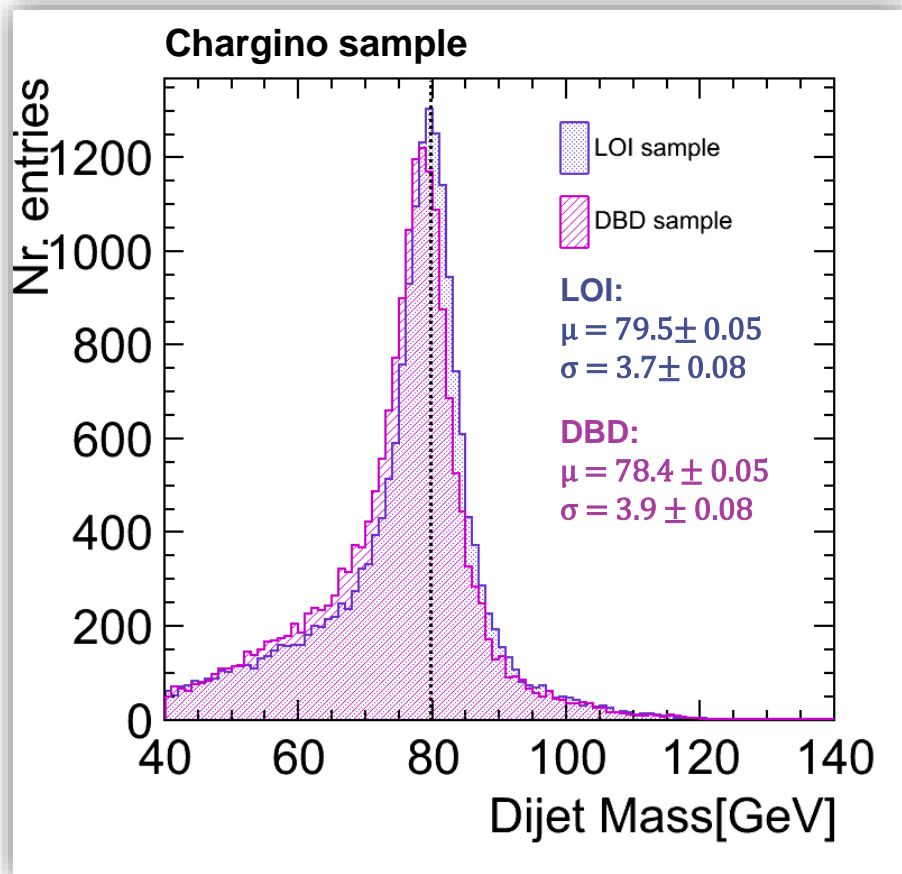
➤ In order to increase performance:

**optimise the number of requested jets and the R-value!**



# Dijet [Boson] Mass Comparison – LOI to DBD

➤ Use dijet mass to separate  $\tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  events → measure cross section



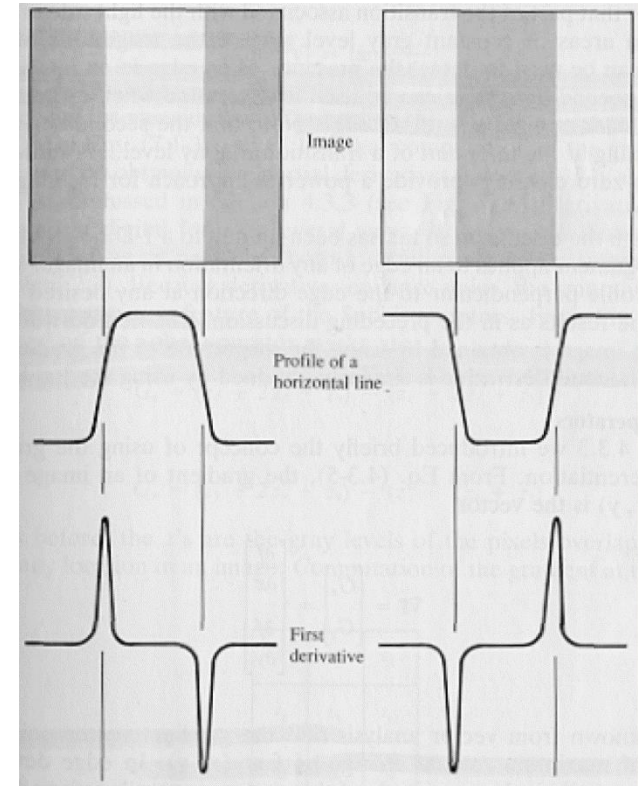
➤ The DBD distribution appears slightly narrower and shifted towards lower energy, however the DBD and LOI distributions are compatible with each other.

# Applying an FIR Filter – Example: the box function

- > The changes of a function can be described by the derivative → interpret the histogram as a 1D function
- > The points that lie on the edge of the distribution → detected by local maxima and minima of the first derivative

$$f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h} \approx f(x+1) - f(x) \quad (h = 1)$$

- > The first derivative is approximated by using the **kernel [-1, 0, 1]**



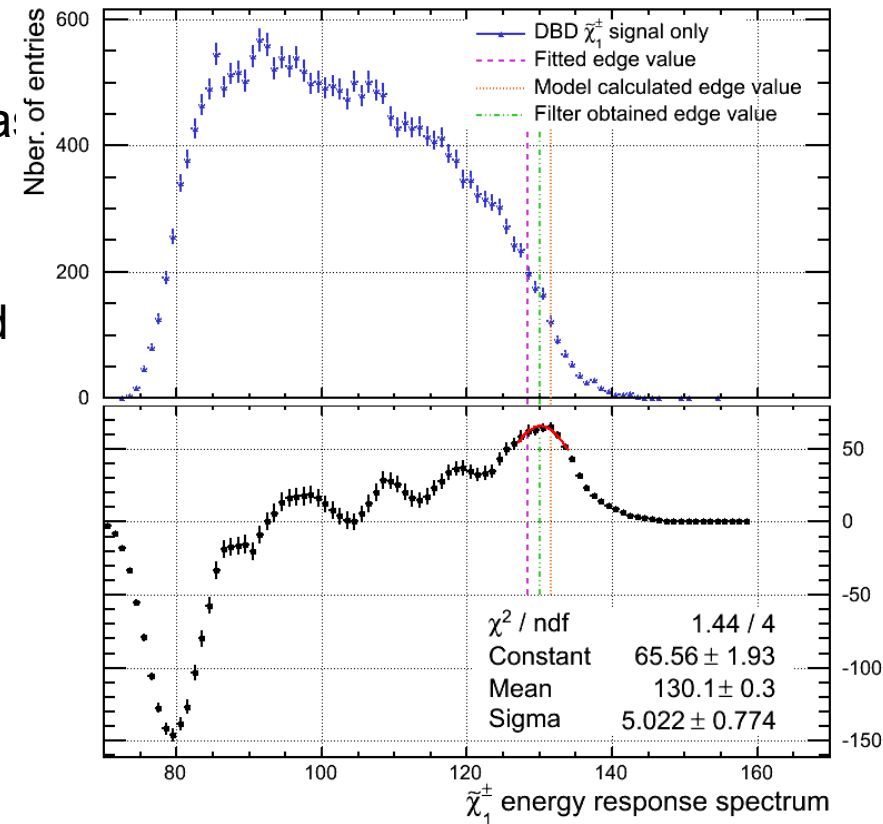
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- > The first derivative is approximated by using the **kernel [-1, 0, 1]**
- > The kernel is convoluted with the histogram:

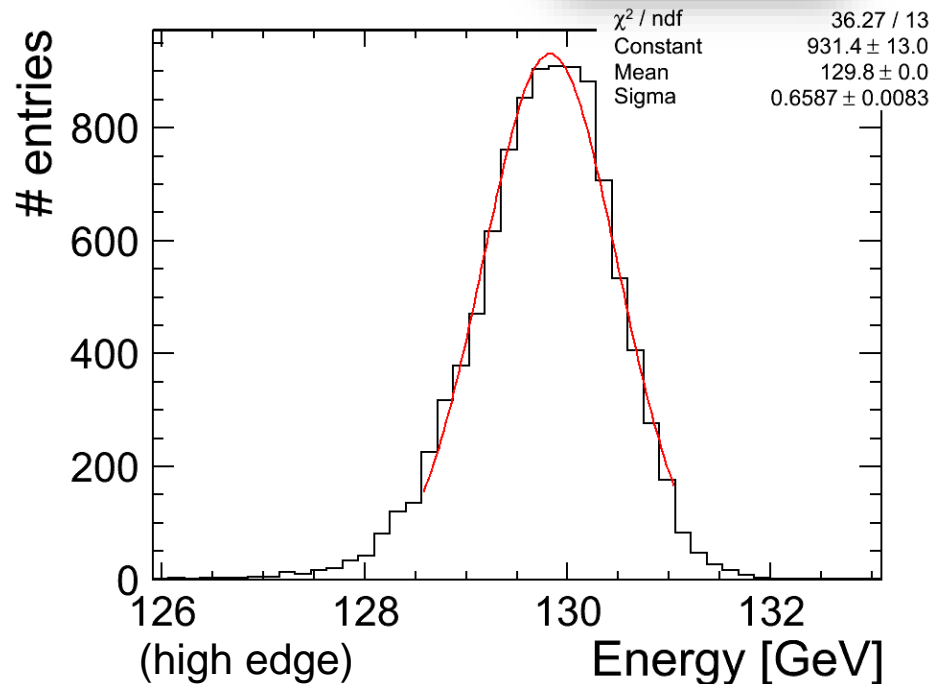
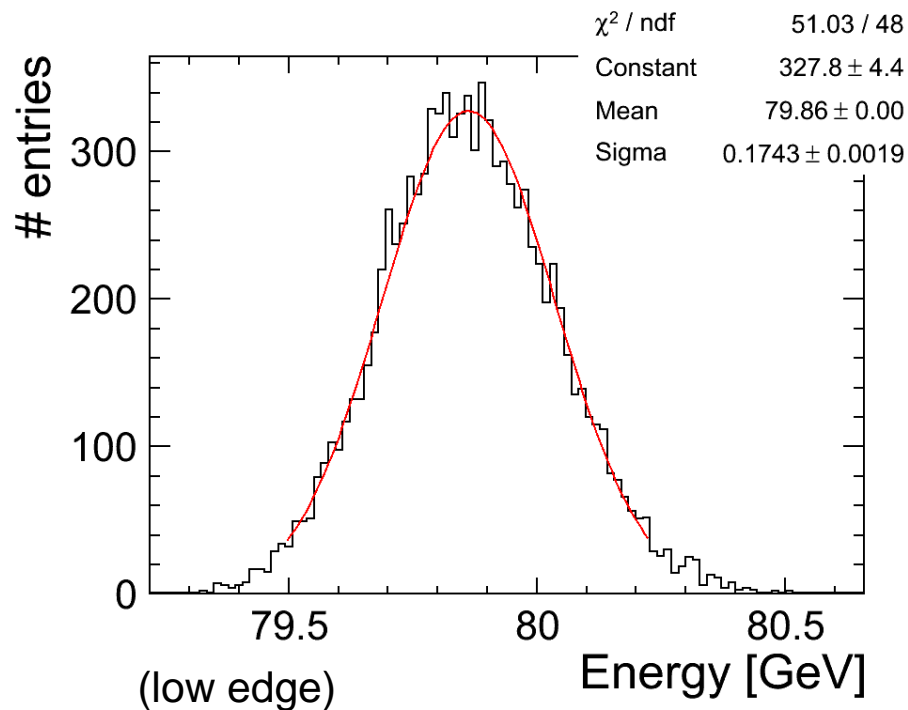
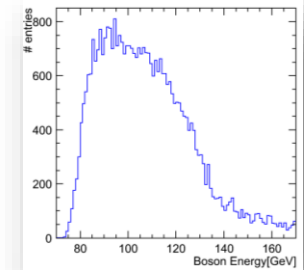
$$response_i = -1 \times bin_{i-1} + 0 \times bin_i + 1 \times bin_{i+1}$$





# Toy MC for the Filter Edge Extraction

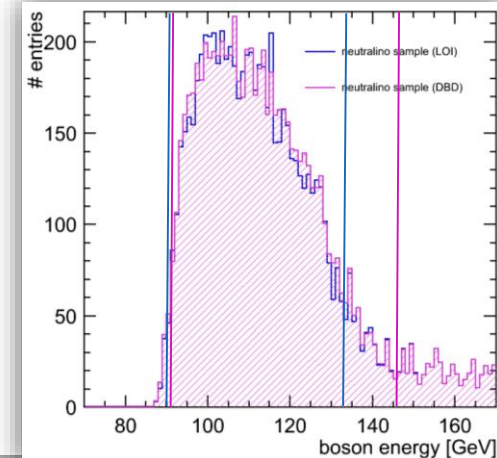
- To estimate the statistical precision of the edge extraction → toy MC
- 10000  $\tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  energy spectra have been produced
- The FDOG filter was then applied 10000 times
- Example: for the  $\tilde{\chi}_1^\pm$  case:



# Issues of the „Endpoint Method“

FCN=59.9958 FROM MIGRAD STATUS=CONVERGED 813 CALLS 815 TOTAL  
EDM=2.74963e-05 STRATEGY= 1 ERROR MATRIX UNCERTAINTY 1.5 per cent

| EXT | PARAMETER | PARABOLIC    | MINOS ERRORS |          |          |
|-----|-----------|--------------|--------------|----------|----------|
| NO. | NAME      | VALUE        | ERROR        | NEGATIVE | POSITIVE |
| 1   | p0        | 7.25426e+01  | 2.24546e+01  |          |          |
| 2   | p1        | 2.13268e+01  | 1.04688e+01  |          |          |
| 3   | p2        | 0.00000e+00  | fixed        |          |          |
| 4   | p3        | -1.06699e+00 | 2.46836e-01  |          |          |
| 5   | p4        | 9.12861e+01  | 6.03729e-01  |          |          |
| 6   | p5        | 1.46148e+02  | 4.78189e+00  |          |          |
| 7   | p6        | 3.49626e-01  | 1.53170e+00  |          |          |
| 8   | p7        | 8.51573e+00  | 2.24546e+00  |          |          |
| 9   | p8        | 3.41940e-01  | 1.25555e+00  |          |          |
| 10  | p9        | 9.12000e+01  | fixed        |          |          |



| Sim. | Edge $W_{\text{low}}$ [GeV] | Edge $W_{\text{high}}$ [GeV] | Edge $Z_{\text{low}}$ [GeV] | Edge $Z_{\text{high}}$ [GeV] |
|------|-----------------------------|------------------------------|-----------------------------|------------------------------|
| DBD  | $79.5 \pm 0.5$              | $130.2 \pm 1.1$              | $91.3 \pm 0.6$              | $146.1 \pm 4.8$              |
| LOI  | $79.7 \pm 0.3$              | $131.9 \pm 0.9$              | $91.0 \pm 0.7$              | $133.6 \pm 0.5$              |

The fitting method appears to be highly dependent on small changes in the fitted distribution → it is clearly NOT appropriate for a comparing the simulation and reconstruction performance.

**We need to apply a different edge extraction method!**

