

# Gaugino Properties Determination in the Fully Hadronic Decay Mode at the ILC

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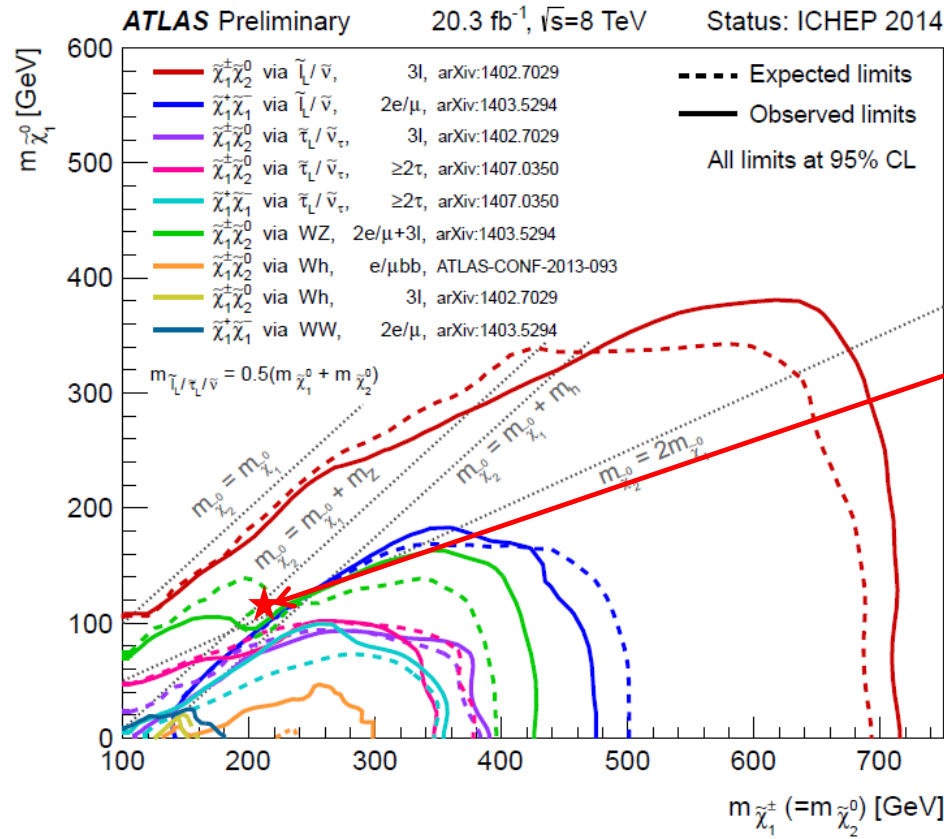


# Study case: $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_2^0$ Pair Production at the ILC

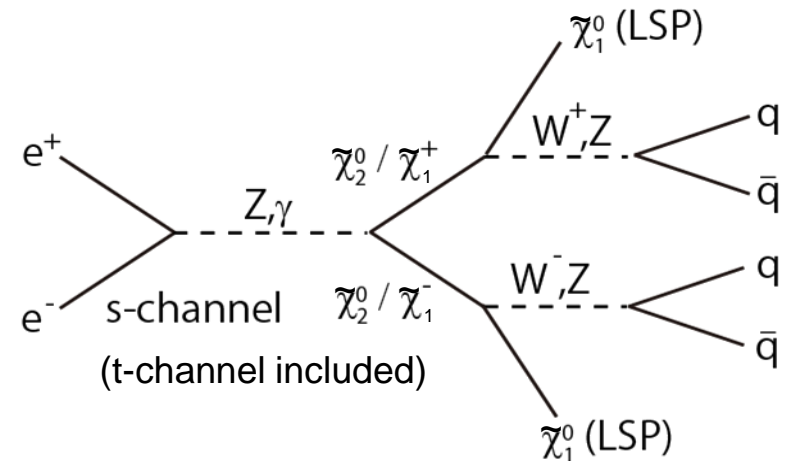
“Point 5” benchmark : gaugino pair production at ILC

<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



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

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



# Analysis Strategy

- > Remove  $\gamma\gamma \rightarrow \text{hadrons}$  background: applied  $k_T$  exclusive algorithm  $\leftrightarrow$  6 jets,  $R=1.1$  (FastJet)
- > Cluster event into 4 jets (Durham)
- > Run kinematic fit (equal mass constraint:  $M_{jj1} = M_{jj2}$ )
  - └ choose jet pairing with best fit probability
- > Run isolated lepton finder (J. Tian and C. Dürig)
- > Perform SUSY selection (12/16 cuts  $\rightarrow$  see [back-up slide](#) )

Sample		$\tilde{\chi}_1^\pm$ hadronic	$\tilde{\chi}_2^0$ hadronic
Selection for mass	Efficiency	90.8% $\rightarrow$ 53%	91% $\rightarrow$ 30%
	Purity	14.7% $\rightarrow$ 63%	2.6% $\rightarrow$ 38%
	Efficiency	72%	73%
	Purity	27%	5%

Selection for x-section



# Mass Measurements



# Gaugino Mass Measurement

- Mass difference to LSP ( $\tilde{\chi}_1^0$ ) is **larger** than  $M_Z \rightarrow$  decays of **real** gauge bosons
- This is a **two-body decay** (well known kinematics!)

- In the gaugino C.M frame:  $(E, p \text{ conservation})$

$$\mathbf{P}_\chi = \mathbf{P}_V + \mathbf{P}_{LSP} \Rightarrow \mathbf{P}_{LSP} = \mathbf{P}_\chi - \mathbf{P}_V$$

$$\text{where } \mathbf{P}_\chi = (M_\chi, \vec{0})$$

$$M_{LSP}^2 = M_\chi^2 + M_V^2 - (2E_\chi E_V - \vec{p}_V \vec{p}_\chi)$$

$$E_V = (M_\chi^2 + M_V^2 - M_{LSP}^2) / 2M_\chi \quad (\text{boson energy})$$

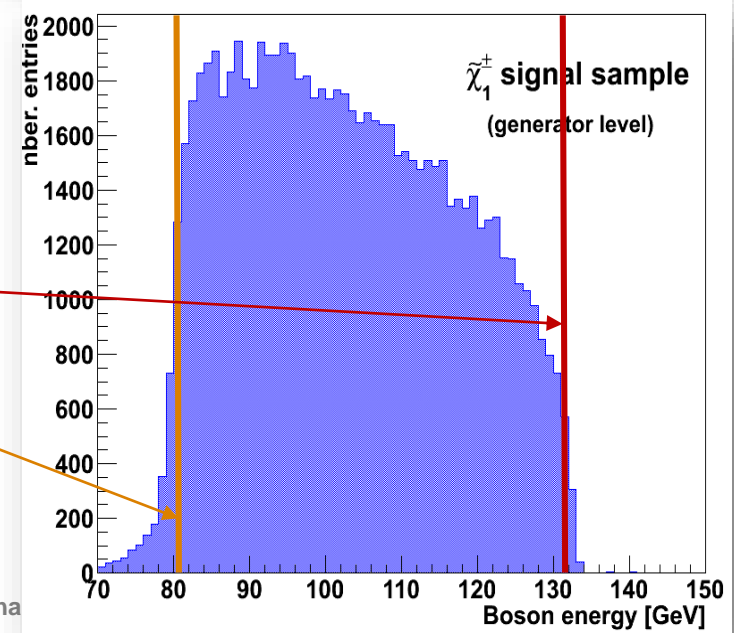
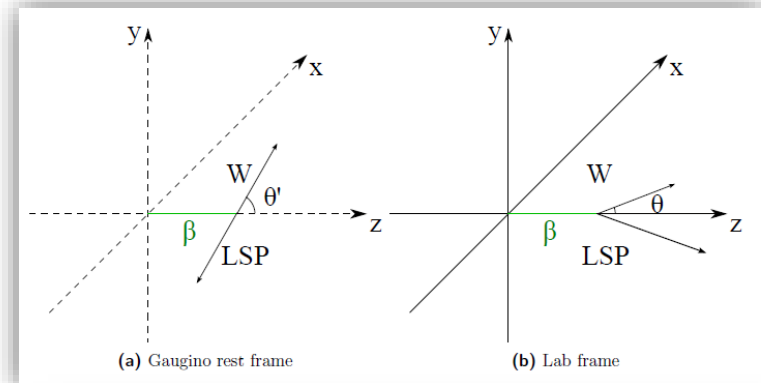
- Boosting into the lab frame:

$$\begin{aligned} E_V^{lab} &= \gamma E_V + \beta \gamma \vec{p}_{V,\parallel} \\ &= \gamma E_V + \beta \gamma |\vec{p}_V| \cos \theta' \end{aligned}$$

$$\theta' = 0 \rightarrow E_V^{lab} = \gamma E_V + \beta \gamma \sqrt{E_V^2 - M_V^2}$$

$$\theta' = \pi \rightarrow E_V^{lab} = \gamma E_V - \beta \gamma \sqrt{E_V^2 - M_V^2}$$

- **Use edge values to calculate gaugino masses!**
- **Two** different **strategies** for edge detection



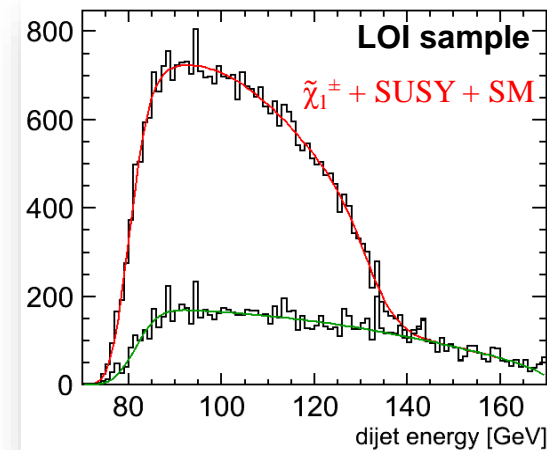
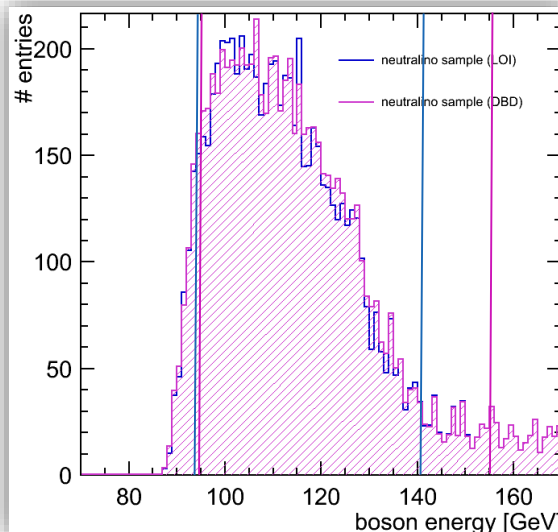
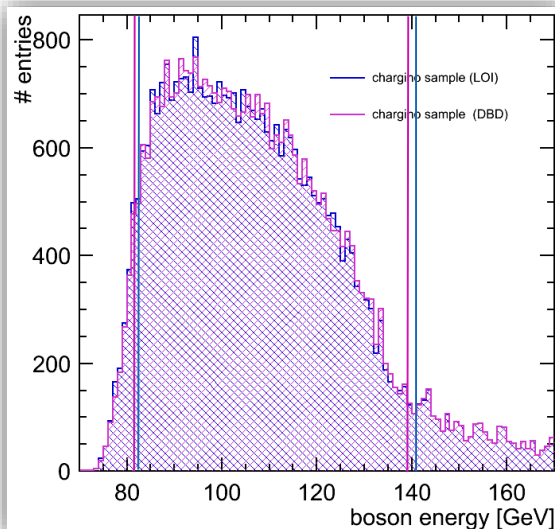
# LOI Strategy: Fit the Boson Energy Spectrum

- Fit dijet energy spectrum and obtain edge positions:

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

- The only free fit parameters: the edge positions  $t_0$  and  $t_1$
- Polynomial → Spectrum slope
- Voigt function → detector resolution and gauge boson width

- Issues with the LOI method:



**Fit method highly sensitive to small fluctuations in energy distribution.**

**Apply a different edge extraction method!**



# DBD Strategy: 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.

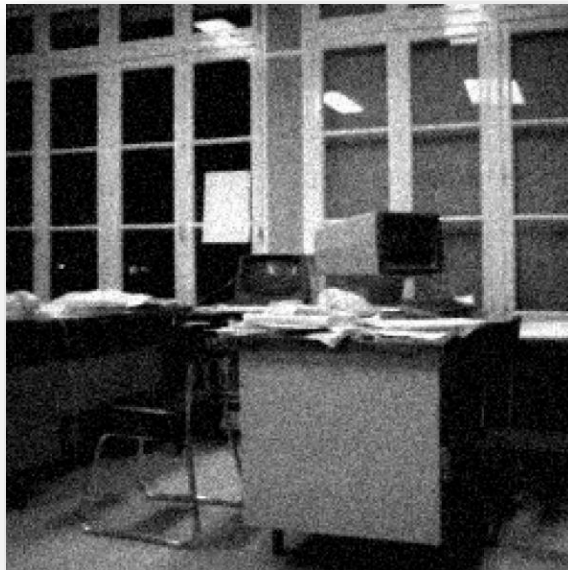
$$y[n] = \sum_{k=-M_1}^{M_2} b_k x[n-k]$$

the input signal

the filter coefficients (weights)

- $y$  is obtained by convolving the input signal with the (finite) weights

D. Demigny, T. Kamlé



Madalina

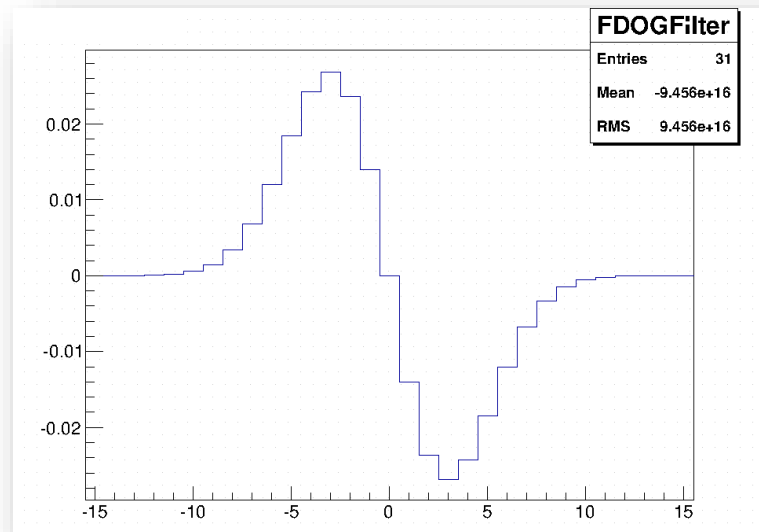


# Choosing the Appropriate Filter

## > Canny's criteria for an optimal filter:

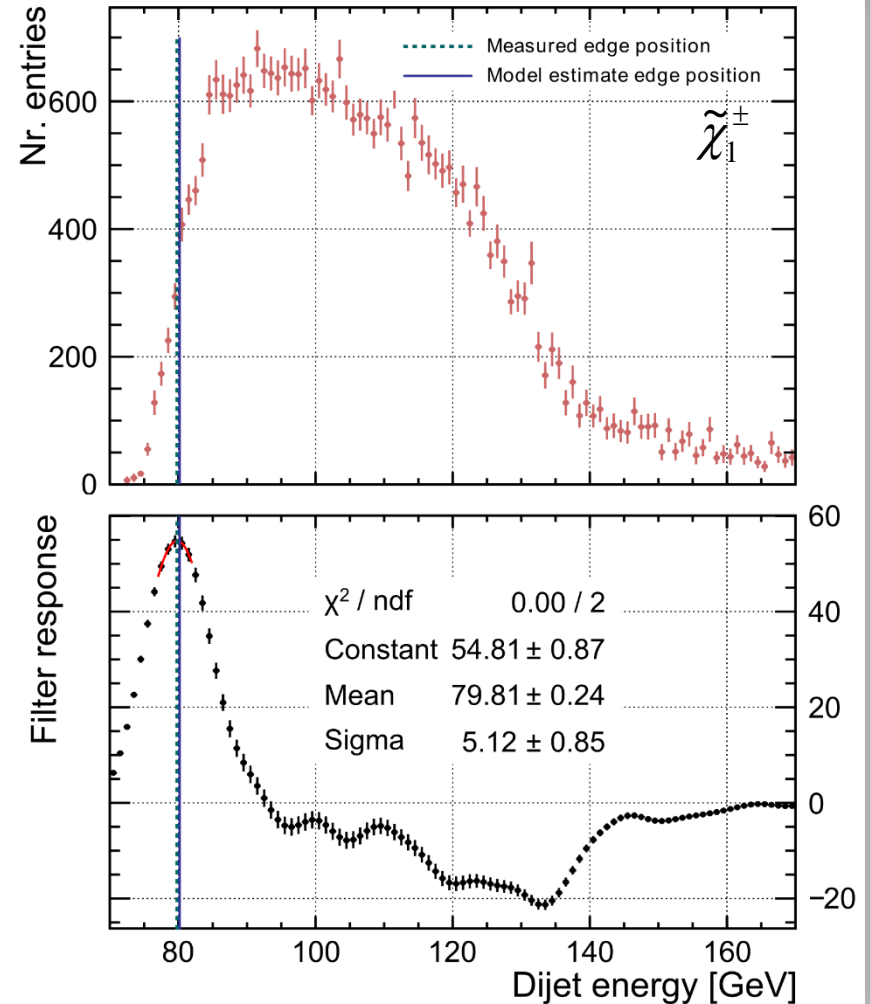
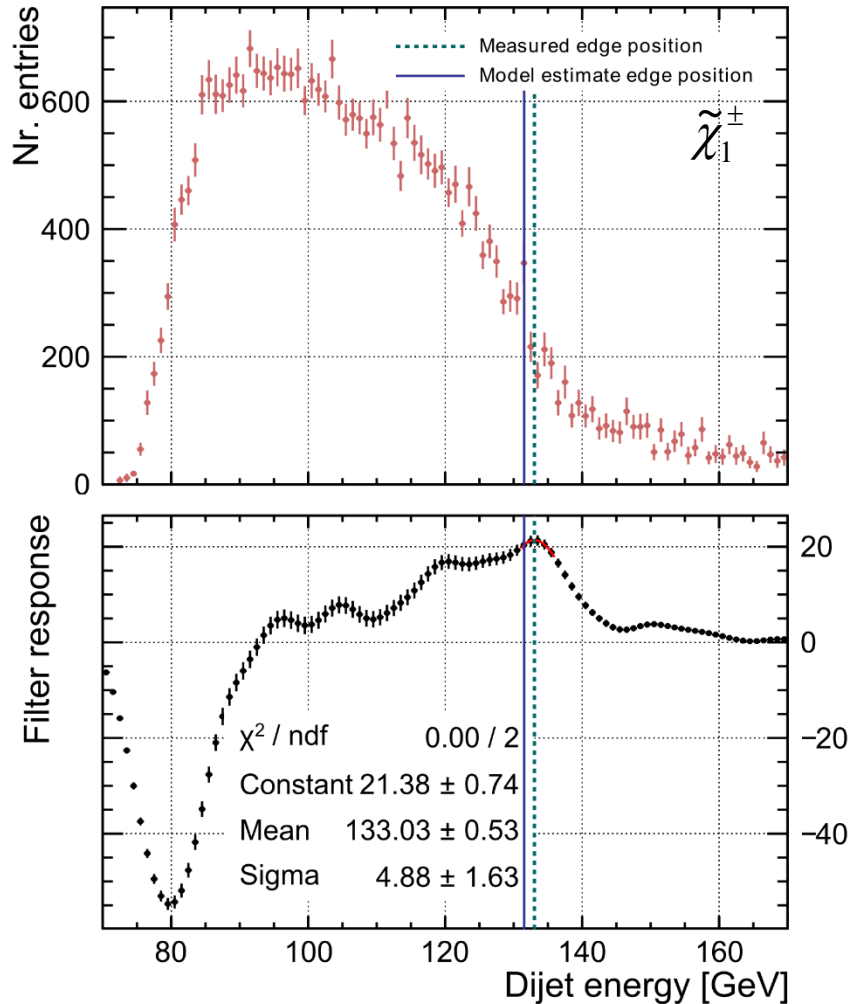
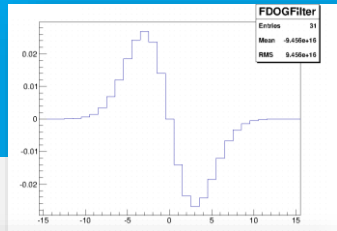
- **J. F. Canny. A computational approach to edge detection.**  
*IEEE Trans. Pattern Analysis and Machine Intelligence*, pages 679-698, 1986
- **Good detection:** probability of obtaining a peak in the response must be high
- **Localisation:** standard deviation of the peak position must be small
- **Multiple response minimisation:** probability of false positive detection must be small

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





# Applying the FIR Filter on DBD Data: Results



# 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]
LOI	80.4±0.2	129.9±0.7	92.3±0.4	128.3±0.9
DBD	79.6±0.2	130.1±0.8	92.1±0.3	128.9±0.8

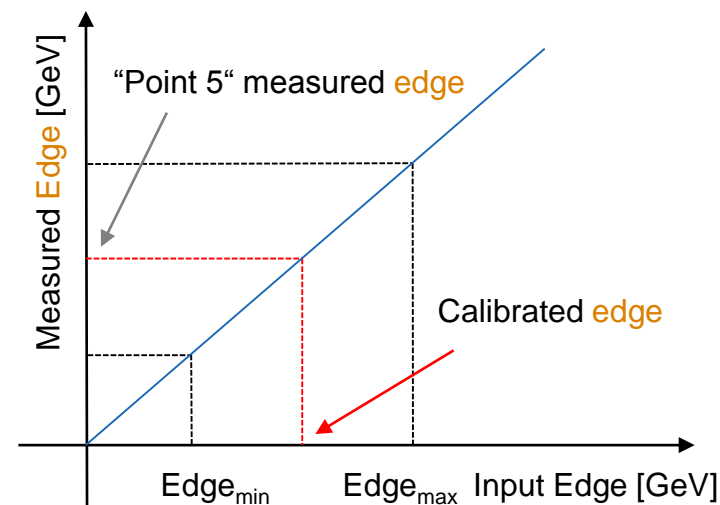
Sample	Mass $\tilde{\chi}_1^\pm$ [GeV]	Mass $\tilde{\chi}_2^0$ [GeV]	Mass $\tilde{\chi}_1^0$ [GeV]
<b>TRUE</b>	<b>216.5</b>	<b>216.7</b>	<b>115.7</b>
LOI	216.9±3.2	220.0±1.4	118.4±1.1
DBD	216.8±3.2	220.6±1.2	118.2±0.9

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



# Edge Calibration → Mass Calibration

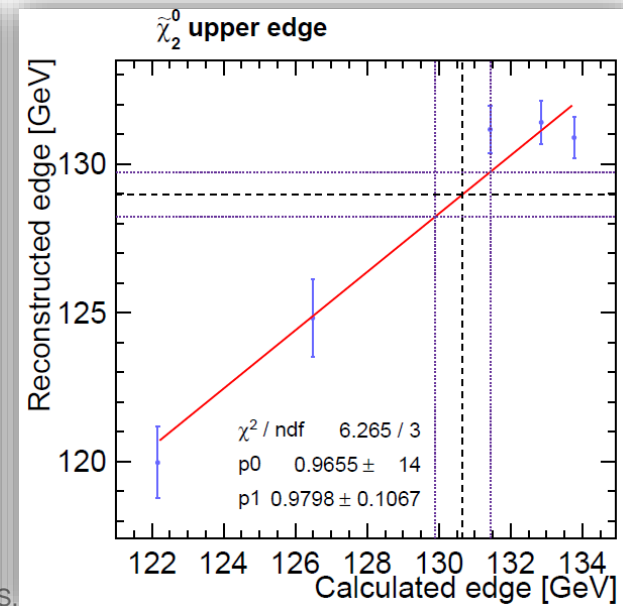
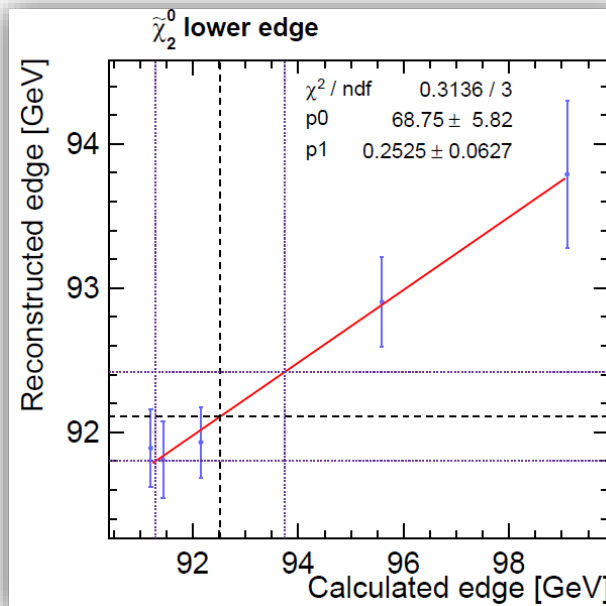
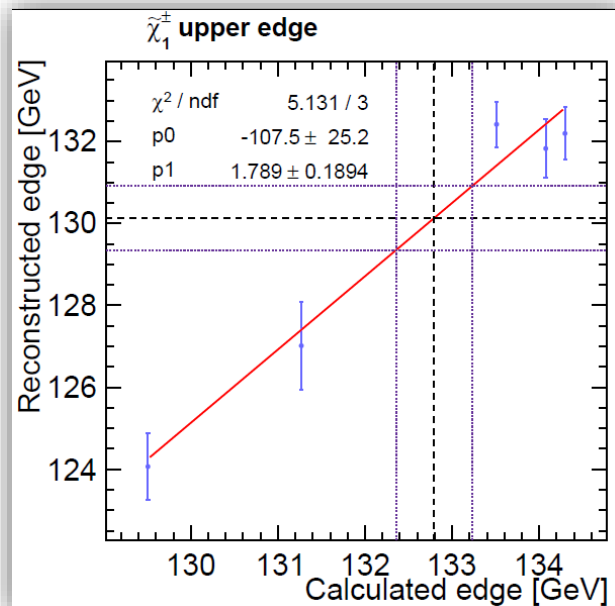
- Performed **only** for **DBD** sample → account for systematics
- Calibrate the edge positions → then calculate the calibrated mass(es)
- Edge calibration procedure:
  - Vary input masses:  $\tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  varied **simultaneously**, **LSP** mass **fixed**!  
 $M_\chi^{min}=210$  GeV  $\leftrightarrow$   $M_\chi^{max}=225$  GeV, 3 GeV step
  - Measure edges for each mass sample  
└→ Obtain **calibration curve**
- Generate the **same number of signal**  $\tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  **events** for all samples
- The **SM background** is the **same** for all mass samples



# Edge Calibration Results I

## ➤ Three different aspects:

1. Calibrate **edges** measured **on generator level** w.r.t. **calculated edges**  
→ study effects of ISR emission, beamstrahlung [0.8% → 1.8%]
2. Calibrate **edges** measured **on reconstruction level** w.r.t. **generator level edges**  
→ study simulation and reconstruction effects [0.2% → 0.9%]
3. Calibrate **edges** measured **on reconstruction level** w.r.t **calculated edges**  
→ take all the effects into account



# Edge Calibration II

## ➤ Three different aspects:

1. Calibrate **edges** measured **on generator level** w.r.t. **calculated edges**  
└→ **study effects of ISR emission, beamstrahlung** [0.8% → 1.8%]
2. Calibrate **edges** measured **on reconstruction level** w.r.t. **generator level edges**  
└→ **study simulation and reconstruction effects** [0.2% → 0.9%]
3. Calibrate **edges** measured **on reconstruction level** w.r.t. **calculated edges**  
└→ **take all the effects into account** [1.1 → 2%]

Gaugino	Mass w/out cal.	Mass with calib.	LOI Mass	Model Mass
$\tilde{\chi}_1^\pm$	$216.7 \pm 3.1$	$214.1 \pm 4.8$	$220.9 \pm 2.9$	216.5 [GeV]
$\tilde{\chi}_2^0$	$220.4 \pm 1.3$	$216.9 \pm 3.4$	$220.6 \pm 1.7$	216.7 [GeV]
$\tilde{\chi}_1^0$	$118.1 \pm 0.9$	$115.5 \pm 1.8$	$118.9 \pm 1.0$	115.7 [GeV]



# Cross Section Measurement



# Cross Section Determination Method

> Interested in:  $\sigma(e^+e^- \rightarrow \tilde{\chi}_1^+ \tilde{\chi}_1^-) \times \text{BR}(\tilde{\chi}_1^+ \tilde{\chi}_1^- \rightarrow \tilde{\chi}_1^0 \tilde{\chi}_1^0 W^+ W^-)$   
 $\sigma(e^+e^- \rightarrow \tilde{\chi}_2^0 \tilde{\chi}_2^0) \times \text{BR}(\tilde{\chi}_2^0 \tilde{\chi}_2^0 \rightarrow \tilde{\chi}_1^0 \tilde{\chi}_1^0 Z^0 Z^0)$

> Relevant observable: the reconstructed dijet [boson] mass

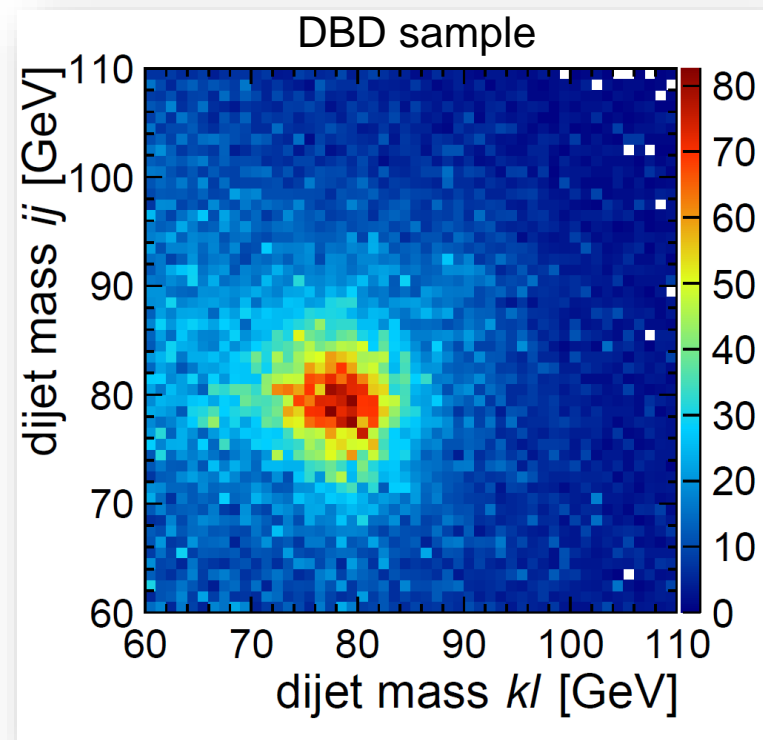
> Relevant distribution: the reconstructed mass of one dijet pair versus the other:

- **AFTER** applying all selection cuts
- Considering only those events for which the kinematic fit has converged
- Including **all** possible dijet associations

> Since  $\sigma \propto \frac{Nr.events}{\varepsilon \cdot \int \mathcal{L}} \Rightarrow$  the goal is to identify the number of  $\tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  events from the total distribution



Perform 2D Template fit.

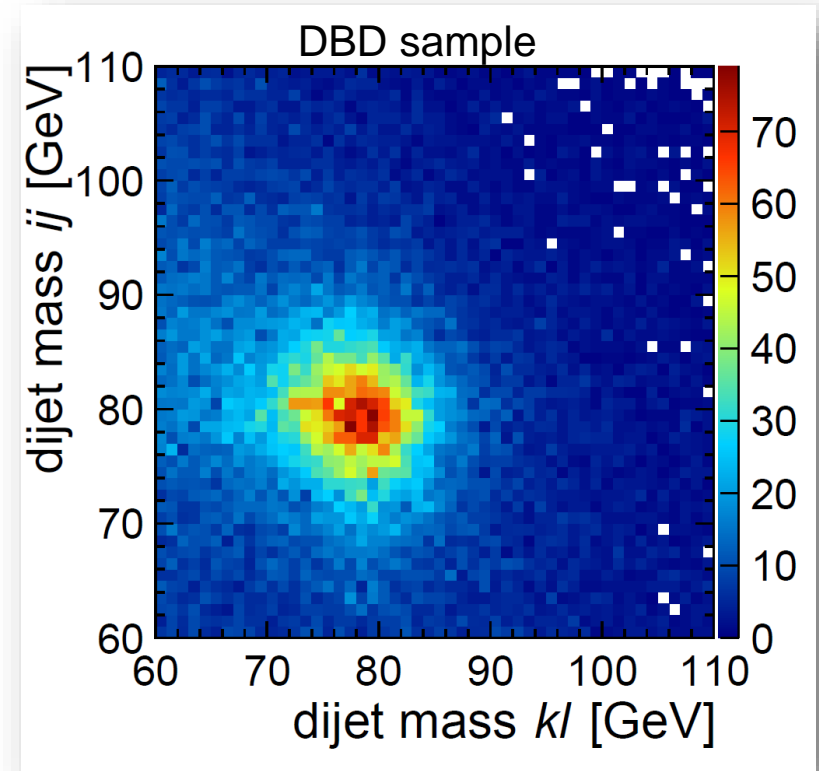


The total distribution (SUSY + SM)

# Cross Section: 2D Template Fit

- Use Monte Carlo data to produce:
  - the chargino template

- After preselection
- Kinematic fit converged
- All dijet permutations included



Chargino events only

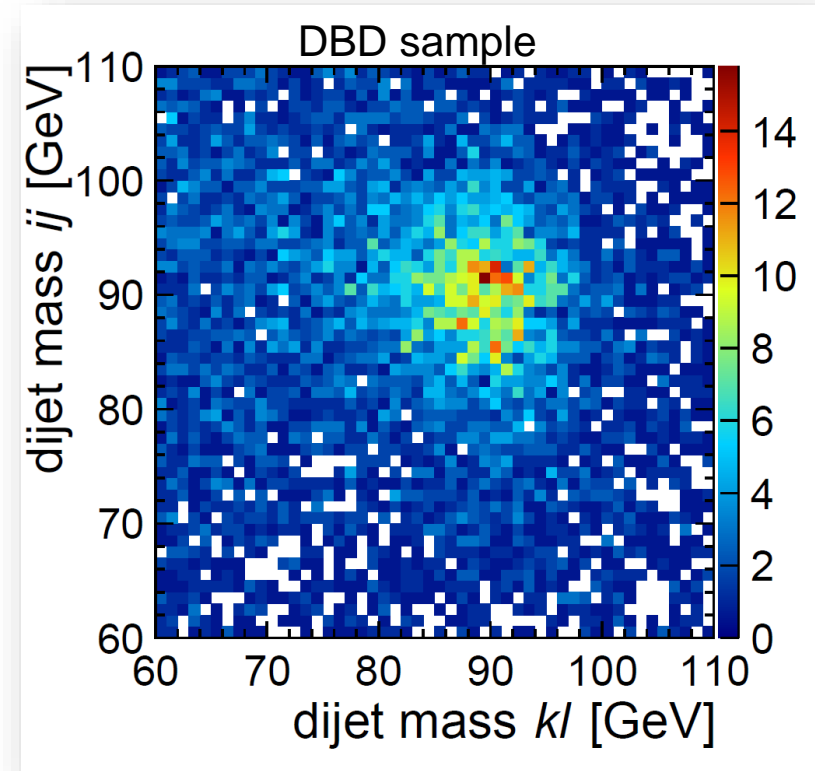


# Cross Section: 2D Template Fit

## > Use Monte Carlo data to produce:

- the chargino template
- the neutralino template

- After preselection
- Kinematic fit converged
- All dijet permutations included



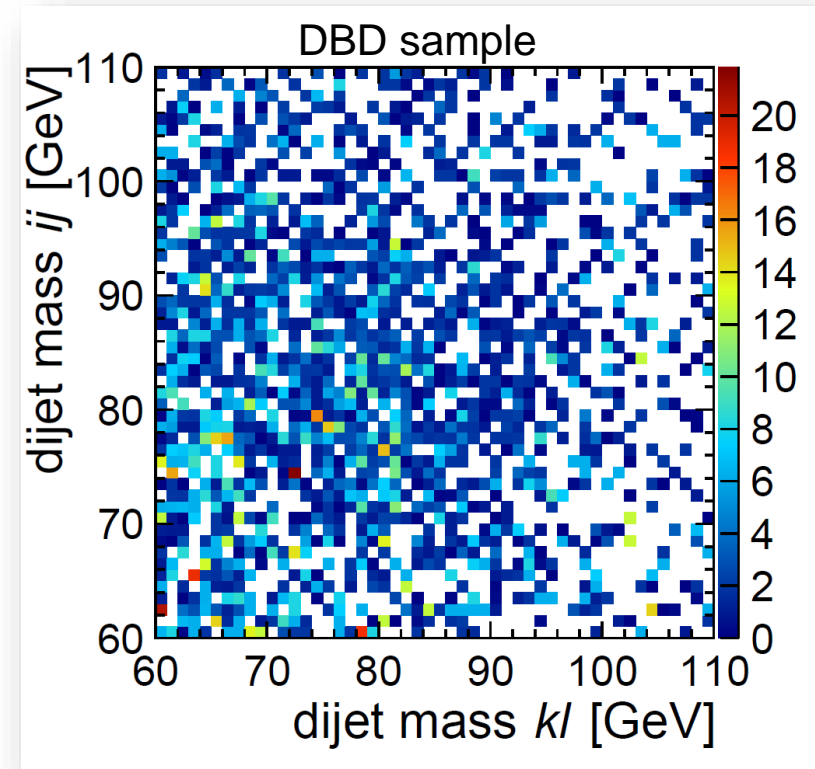
Neutralino events only

# Cross Section: 2D Template Fit

## > Use Monte Carlo data to produce:

- the chargino template
- the neutralino template
- the SM background template

- After preselection
- Kinematic fit converged
- All dijet permutations included



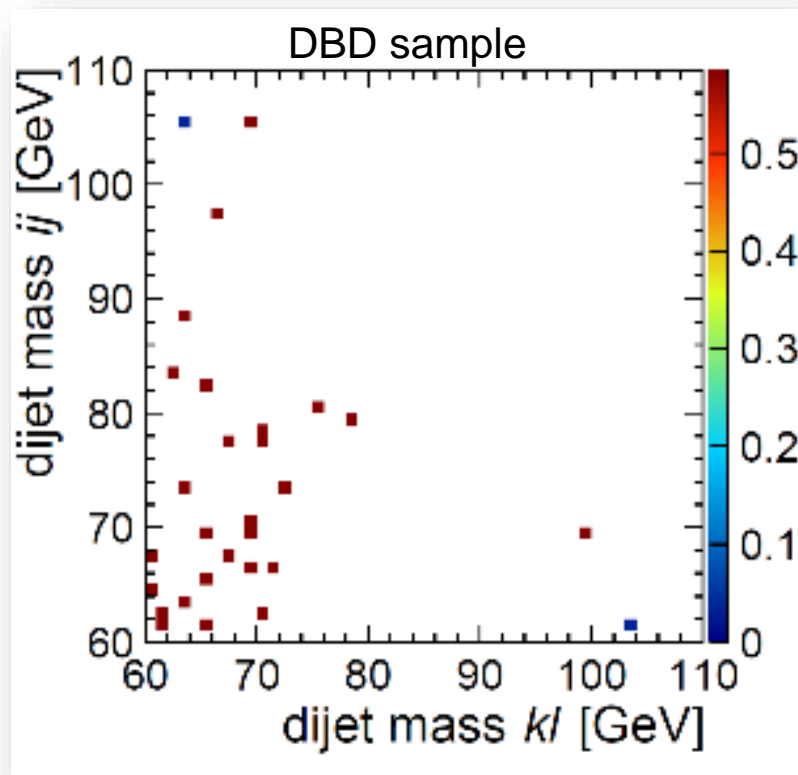
Standard Model events only

# Cross Section: 2D Template Fit

## > Use Monte Carlo data to produce:

- the chargino template
- the neutralino template
- the SM background template
- the SUSY background → **negligible!**

- After preselection
- Kinematic fit converged
- All dijet permutations included

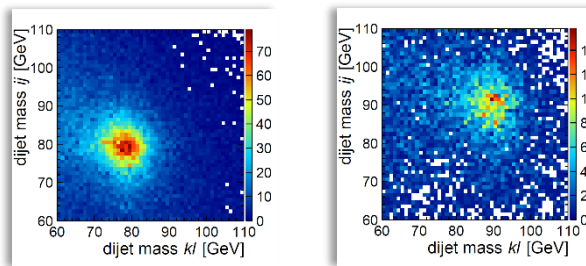


SUSY background events only



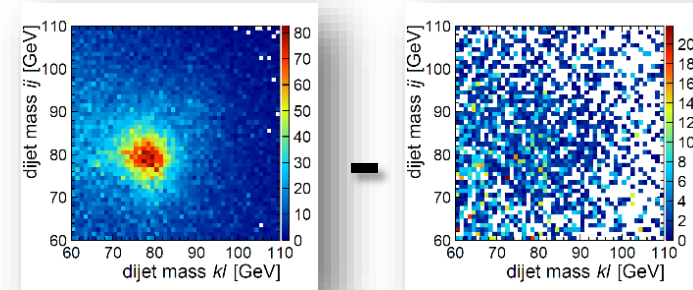
# Cross Section: 2D Template Fit

- The fitting procedure:
- Subtract the SM background template from the total data distribution
- Defining the two-dimensional fitting function:



$$f_{Fit}(x, y) = \overbrace{a \cdot f_{\tilde{\chi}_1^\mp}(x, y)} + \overbrace{b \cdot f_{\tilde{\chi}_2^0}(x, y)}$$

- $a$  and  $b \rightarrow$  the only free parameters
  - $a$  and  $b =$  the fraction of template events found in the total data distribution
  - in an ideal case,  $a = b = 1$
- Apply the template fit on the remaining data events

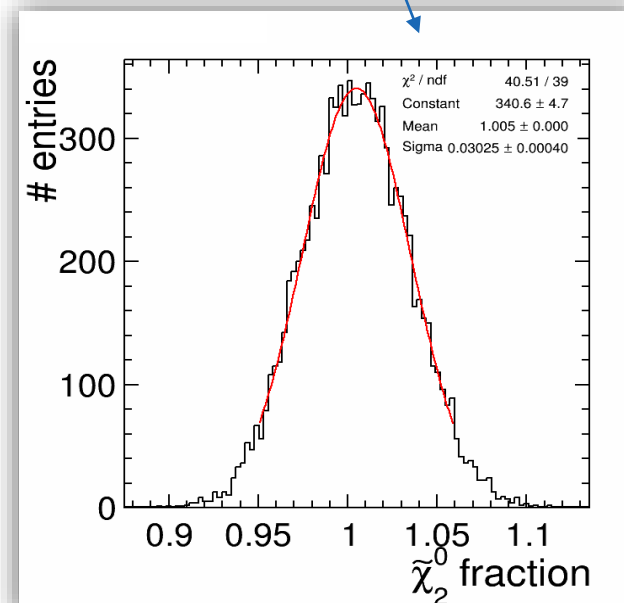
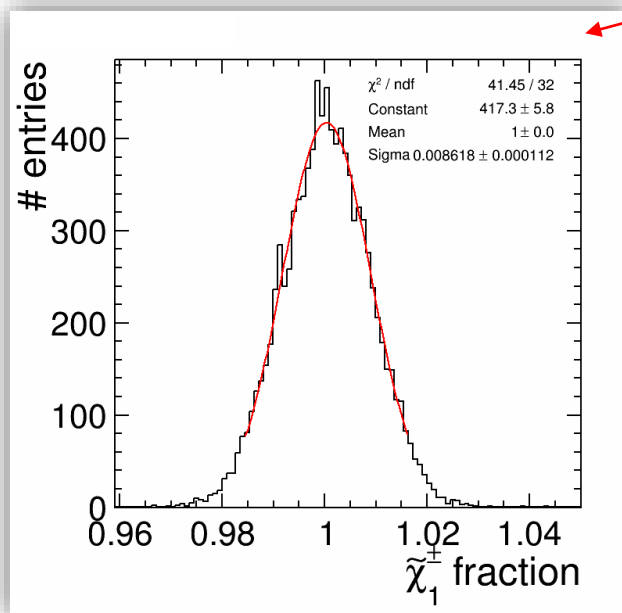


# 2D Template Fit Toy Monte Carlo

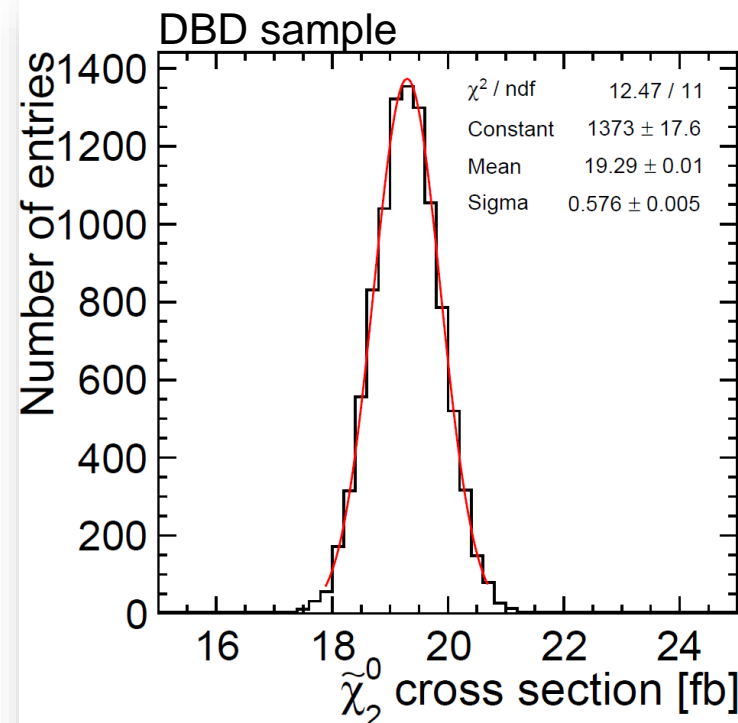
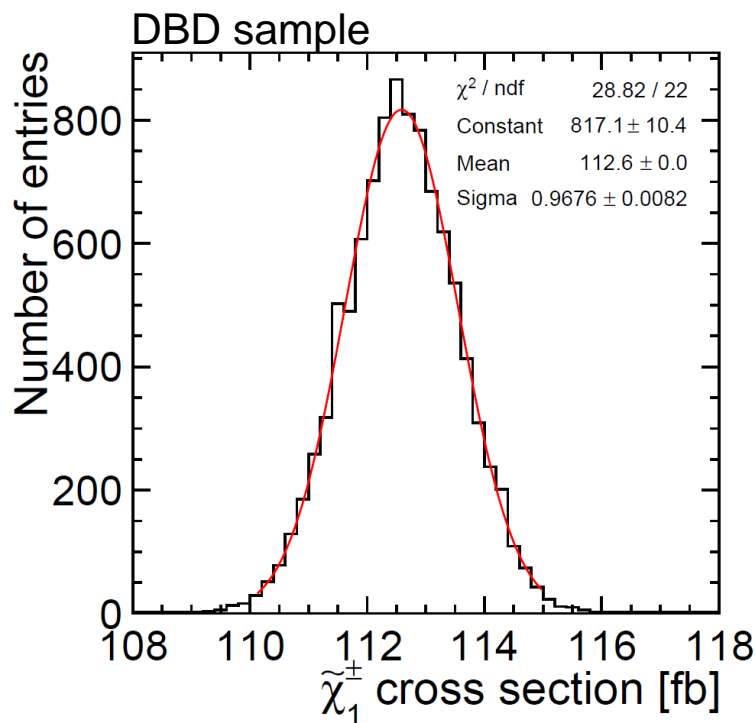
> **Note:** limited amount of Monte Carlo data available → **toy Monte Carlo study**

> **Running the toy MC:**

- Treat the total data distribution as a p.d.f
- Randomly sample the initial distribution  $N$  times:  $N = N_{\text{evts.}}^{\text{initial}} \pm \sqrt{N_{\text{evts.}}^{\text{initial}}}$
- Subtract the SM template from the new distribution
- Apply the fitting function → obtain one value each for  $a$  and  $b$
- Repeat procedure 10000 times



# 2D Template Fit: Results



$$a_{\text{mean}} = 1.00 \pm 0.009$$

$$b_{\text{mean}} = 1.01 \pm 0.03$$

Sample	$\tilde{\chi}_1^\pm$ x- section [fb]	$\tilde{\chi}_2^0$ x-section [fb]
Generator	112.54	19.2
DBD	$112.6 \pm 0.97$	$19.3 \pm 0.58$



# Cross Section: 2D Template Fit – Comparison to LOI

> The same procedure has been applied to the LOI data:

Sample	$\tilde{\chi}_1^+$ x- section [fb]	$\tilde{\chi}_2^0$ x-section [fb]
Generator level	132.2	22.8
LOI	$132.2 \pm 1.1$	$23.2 \pm 0.7$
arXiv:0906.5508v2	$132.9 \pm 0.9$	$22.5 \pm 0.5$

Sample	$\tilde{\chi}_1^+$ x- section [fb]	$\tilde{\chi}_2^0$ x-section [fb]
Generator level	112.5	19.2
DBD	$112.6 \pm 0.97$	$19.3 \pm 0.6$

- **Note** - the difference between cross sections at generator level
  - Difference in beam-spectrum
  - Missing processes - Whizard 1.95



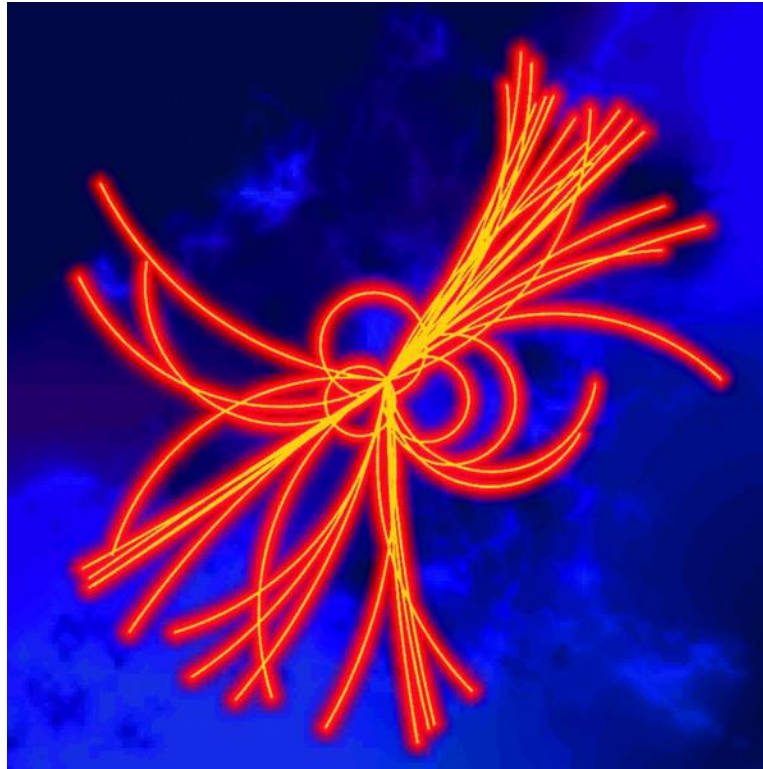
# Conclusions

- The  $\tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  pair production in the framework of the “Point 5” benchmark has been presented as study case.
- **Mass measurements:**
  - LOI fitting method for edge measurement very sensitive to small changes
  - Applying a finite impulse response (FIR) filter instead: more robust (i.e., independent on distribution shape), provides just as good if not better statistical precision.
  - A mass calibration procedure was performed for the DBD sample: **beam related effects twice as large effect as sim. + reco. impact!**
- **Cross section measurements:**
  - A 2D template fitting procedure for cross-section determination was presented.
  - Due to limited amounts of available Monte Carlo data perform a toy Monte Carlo study.
  - Procedure applied both on LOI as well as on DBD data.
  - Mean cross-section values very close to the model values in both cases → cross-check for the procedure performance.
  - Despite increased detector realism and addition of  $\gamma\gamma$  background statistical uncertainties are very similar for both data samples:  $\approx 1\%$  for  $\tilde{\chi}_1^\pm$  and  $\approx 3\%$  for  $\tilde{\chi}_2^0$





# Thank You!



# Study Case - Motivation

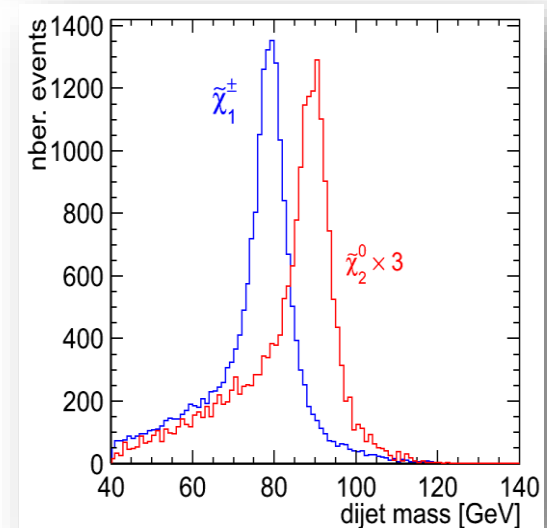
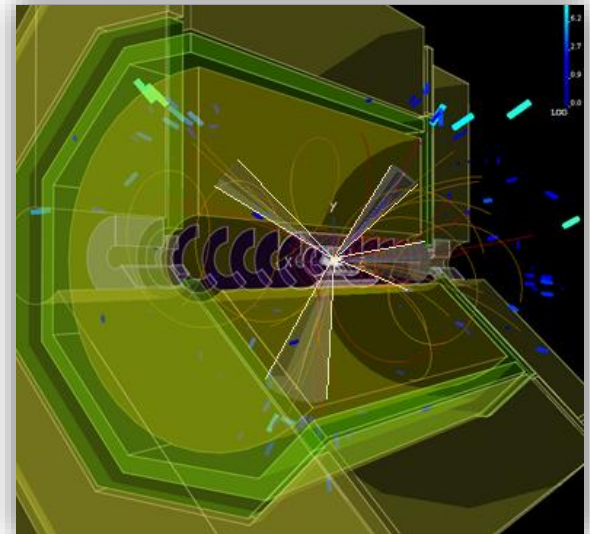
## ➤ Signal topology:

- 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

## ➤ Challenge 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

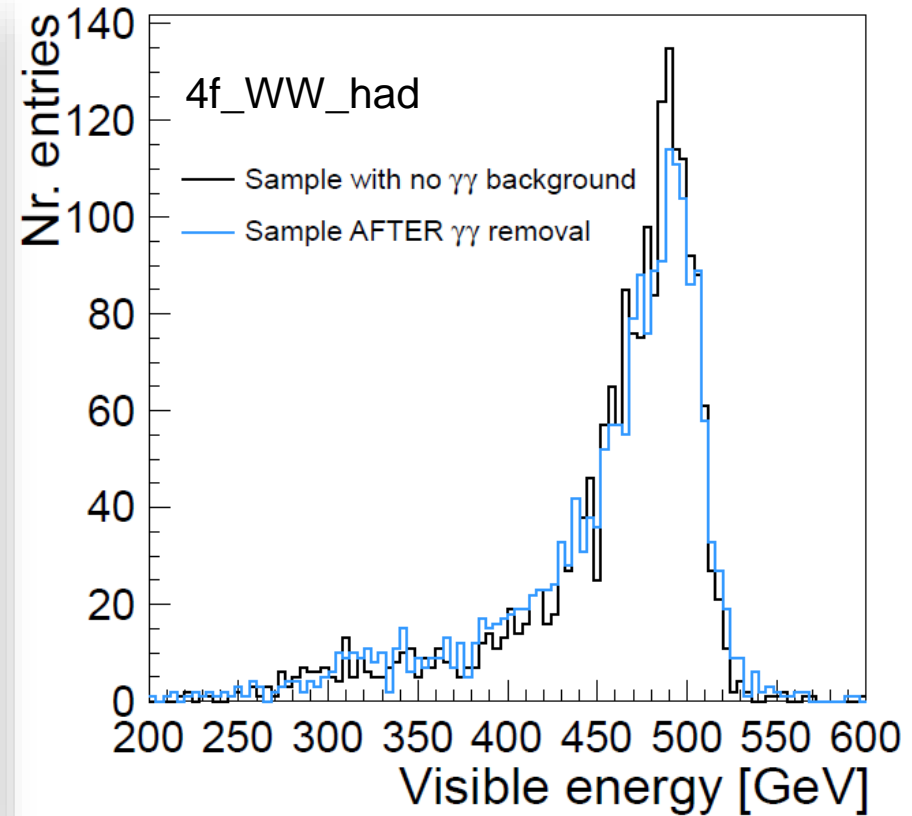
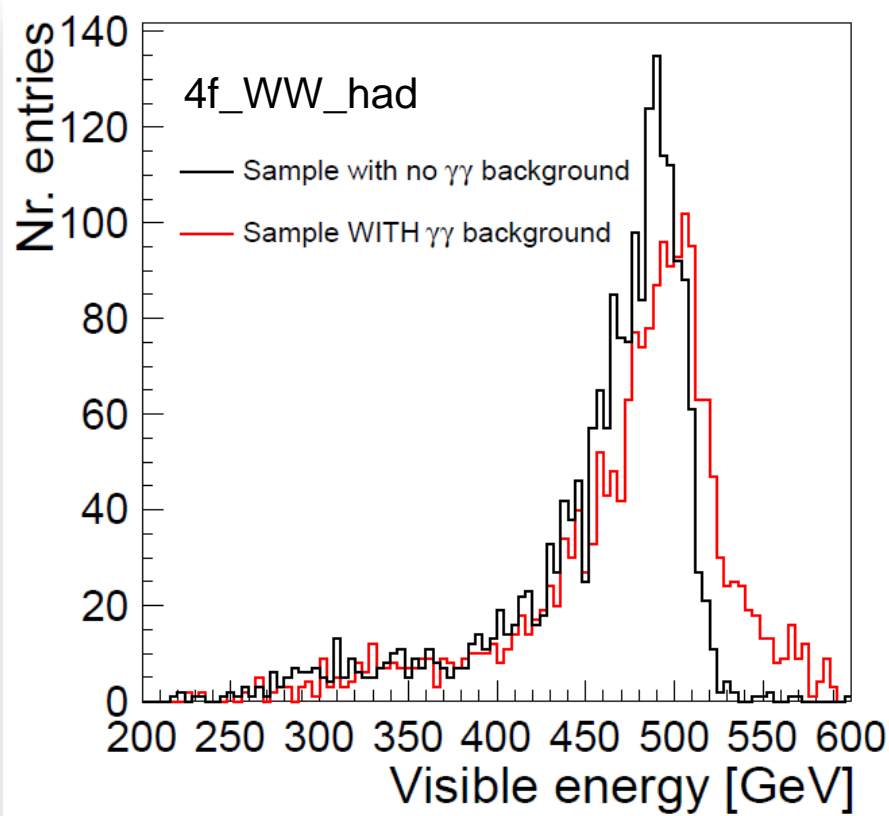
- Signal + background were simulated and reconstructed with `ilcsoft v01-06`
- The jet energy scale was increased by 1%
- No  $\gamma\gamma$  background overlay
- The analysis was run on existing data samples

- Some processes could not be produced in Whizard 1.95
- 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 run



# Analysis Strategy

- Remove  $\gamma\gamma \rightarrow \text{hadrons}$  background: applied  $k_T$  exclusive algorithm  $\leftrightarrow$  6 jets,  $R=1.1$  (FastJet)



Sample	$\tilde{\chi}_1^\pm$ hadronic (signal)	$\tilde{\chi}_2^0$ hadronic (signal)	SUSY background	$\gamma\gamma$ & $\gamma e$ (SM)	2 fermions (SM)	4 fermions (SM)	6 fermions (SM)
No cut	27427	4897	71450	173791	1.1239e+ 07	1.60385e+ 07	589188
No isolated leptons found	27281	4857	39592	51136	1.02105e+ 07	9.21406e+ 06	372419
Nber. PFOs in event	27274	4853	28936	38553	8.61602e+ 06	6.73891e+ 06	311637
Nber. tracks with $P_T > 1$ GeV in event	27228	4851	25530	34803	7.60753e+ 06	6.22246e+ 06	282188
Thrust	27213	4845	24996	34347	4.57776e+ 06	4.9343e+ 06	281913
Nber. tracks in event	27193	4841	23049	33647	4.27554e+ 06	4.81107e+ 06	281652
Visible energy	27159	4831	20935	21830	2.88111e+ 06	926895	17059
Jet energy	27141	4829	17895	21360	2.55856e+ 06	846448	16914
Jet $\cos(\theta)$	26530	4729	15964	17582	1.78384e+ 06	607998	16049
$y_{34}$	26372	4704	11202	16231	330943	299658	14892
Nber tracks in jet	25434	4585	8083	14666	261520	205867	12125
Miss $\cos(\theta)$	25171	4535	8020	4489	9171	117756	11656
Lepton energy	24913	4460	7749	4281	8432	109365	10121
Nber. PFOs in jet	24737	4444	7305	4148	8253	102304	9783
Miss $\cos(\theta)$	19868	3589	6135	1383	1100	53957	6955
Missing mass	19830	3584	6134	1175	931	41326	1764
Kinematic fit converged	19753	3565	5966	1152	839	40263	1749

Blue: selection for the mass measurement

Red: selection for the cross section measurement

# $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_2^0$ Signal Separation

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

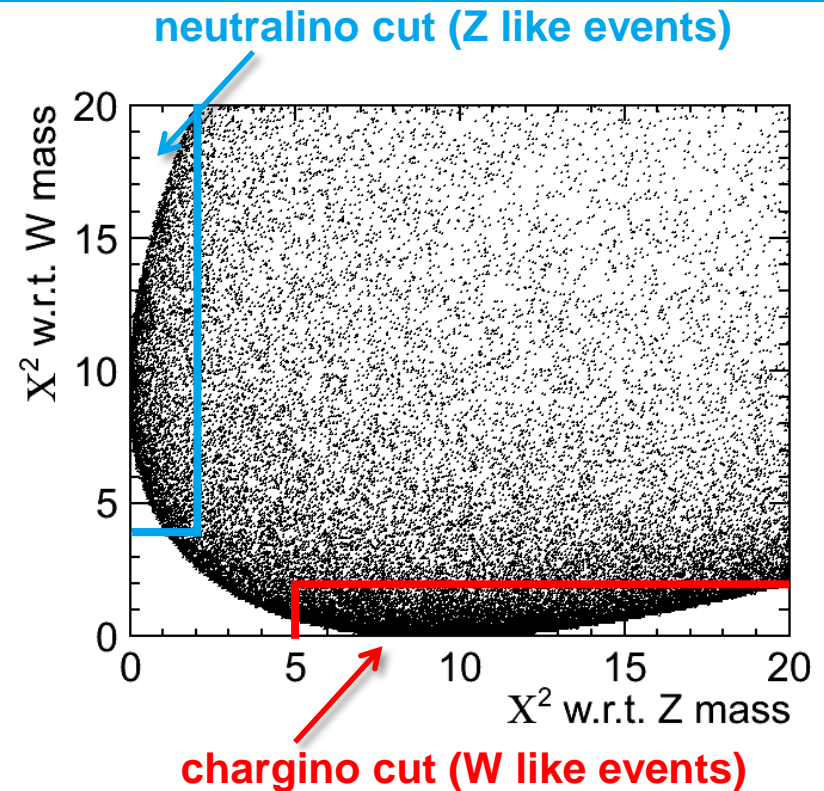
$$\chi^2(m_{j1}, m_{j2}) = \frac{(m_{j1} - m_V)^2 + (m_{j2} - m_V)^2}{\sigma^2}$$



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

- > Downside: lose statistics
  - Cut away 47% of  $\tilde{\chi}_1^\pm$  surviving events
  - Cut away 61% of  $\tilde{\chi}_2^0$  surviving events
- > However, after the  $\chi^2$  cut, the separation is quite clear:

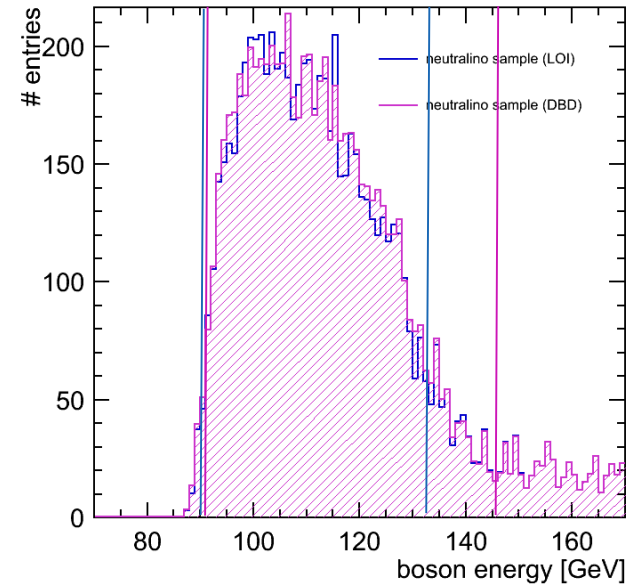
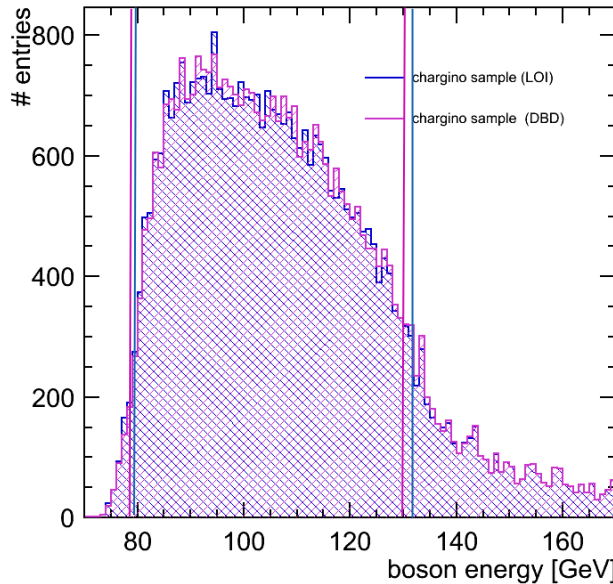
Sample	$\tilde{\chi}_1^\pm$ hadronic	$\tilde{\chi}_2^0$ hadronic
Efficiency	90.8%	91%
Purity	14.7%	2.6%



Obs.	DBD	
	$\tilde{\chi}_1^\pm$	$\tilde{\chi}_2^0$
Efficiency	53%	30%
Purity (total)	63%	38%
Purity (SUSY)	94%	62%



# Issues of the LOI Strategy



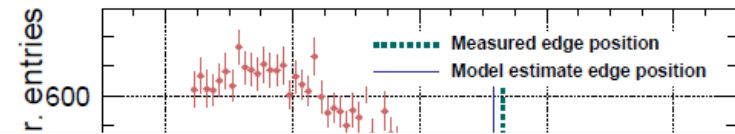
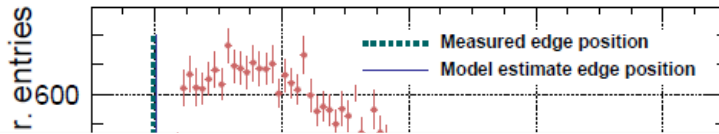
Sim.	Edge $W_{\text{low}}$ [GeV]	Edge $W_{\text{high}}$ [GeV]	Edge $Z_{\text{low}}$ [GeV]	Edge $Z_{\text{high}}$ [GeV]
DBD	$79.5 \pm 0.5$	$130.2 \pm 1.1$	$91.3 \pm 0.6$	$146.1 \pm 4.8$
LOI	$79.7 \pm 0.3$	$131.9 \pm 0.9$	$91.0 \pm 0.7$	$133.6 \pm 0.5$

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!**

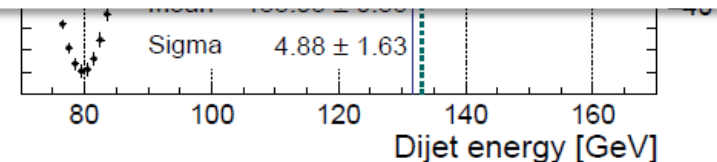
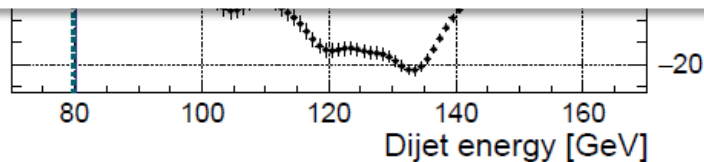


# Applying the FIR Filter on DBD Data: Results



filter

Calc.	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	$79.7 \pm 0.3$	$131.9 \pm 0.9$	$91.0 \pm 0.7$	$133.6 \pm 0.5$
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	$79.6 \pm 0.2$	$130.1 \pm 0.8$	$92.1 \pm 0.3$	$128.9 \pm 0.8$



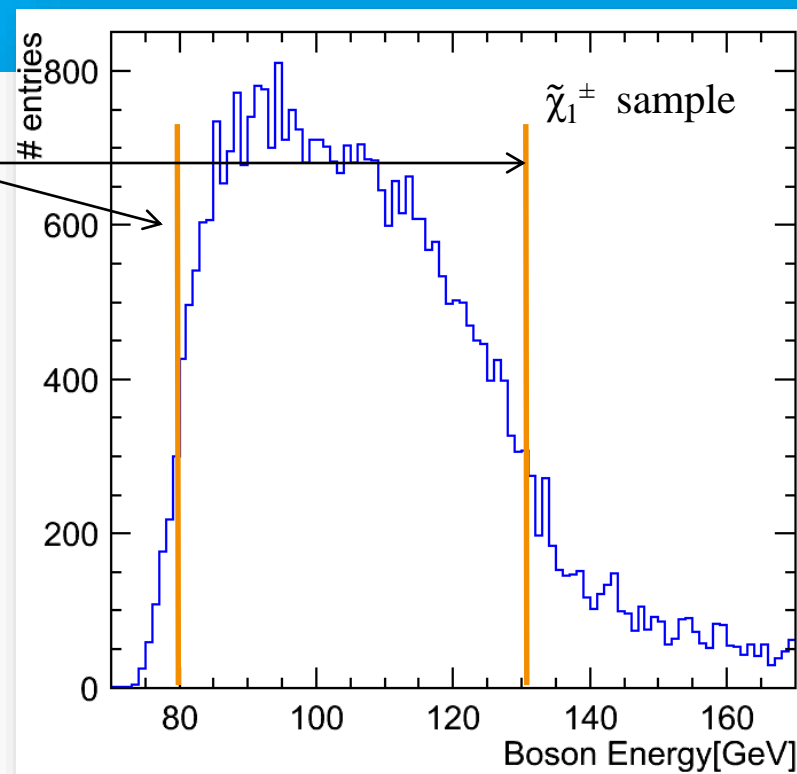
- **Statistical errors determined from toy Monte Carlo**





# Applying an FIR Filter

- > Goal: find edge positions in spectrum



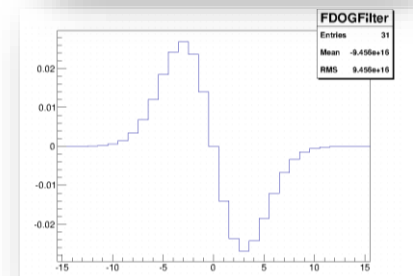
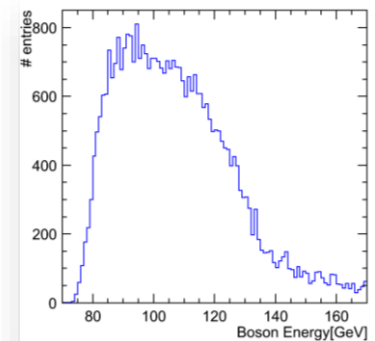
# Applying an FIR Filter

- > Goal: find edge positions in spectrum
- > Strategy:
  - Choose an FIR filter (kernel)
  - 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

Thanks to S. Caiazza.



# Applying an FIR Filter

> Goal: find edge positions in spectrum

> Strategy:

- Choose an FIR filter (kernel)
- Note: filter length  $\ll$  signal histogram length
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- Calculate dot product between **Signal** and **Filter** → **obtain one value**

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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

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> Strategy:

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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

# Applying an FIR Filter

> Goal: find edge positions in spectrum

> Procedure:

- Choose an FIR filter (kernel)
- 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

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

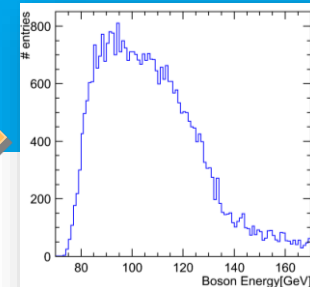
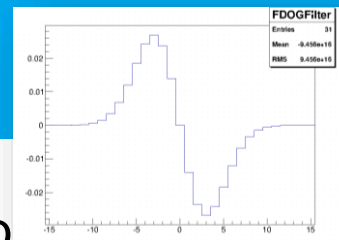
- There are 3 filter parameters that can be optimised
  - The width of the Gaussian ( $\sigma$ )
  - The kernel size (# bins of the filter histogram)
  - The binning of the input boson energy histogram



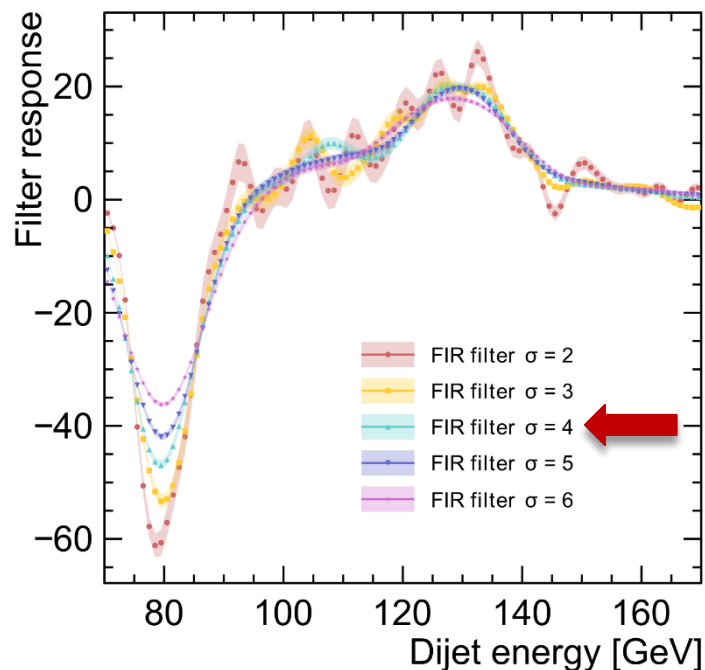
# FDOG Filter Optimisation

➤ There are 3 filter parameters that can be optimised

- The width of the Gaussian ( $\sigma$ )
- The kernel size (# bins of the filter histogram)
- The binning of the input boson energy histogram



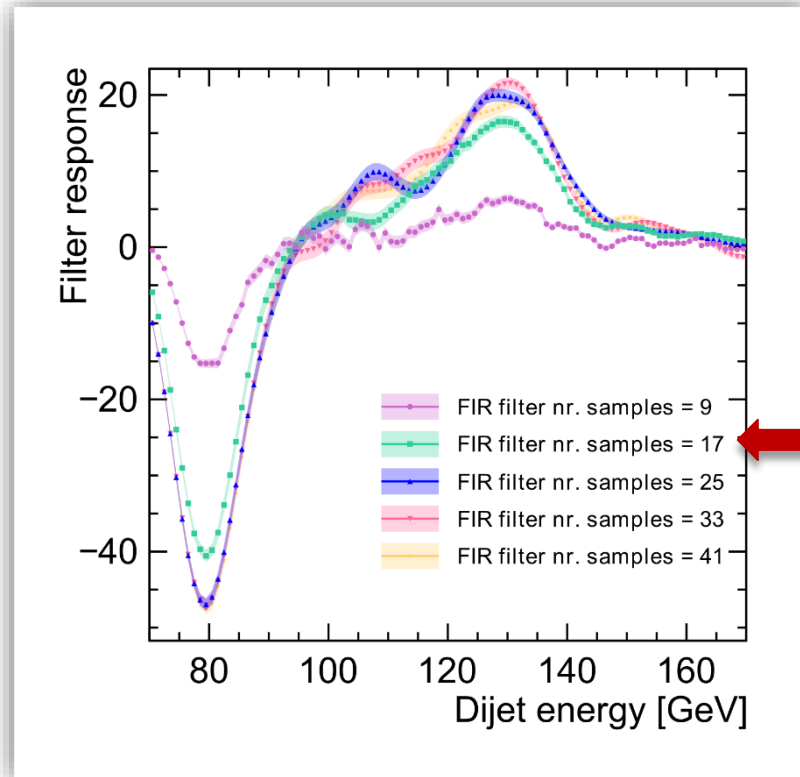
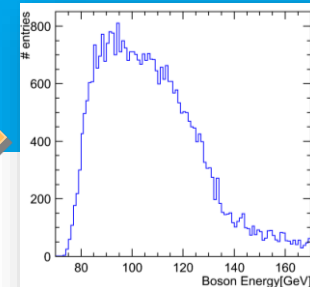
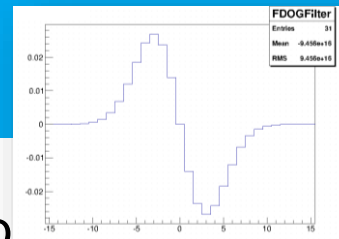
*(the kernel and bin sizes were fixed)*



# FDOG Filter Optimisation

> There are 3 filter parameters that can be optimised

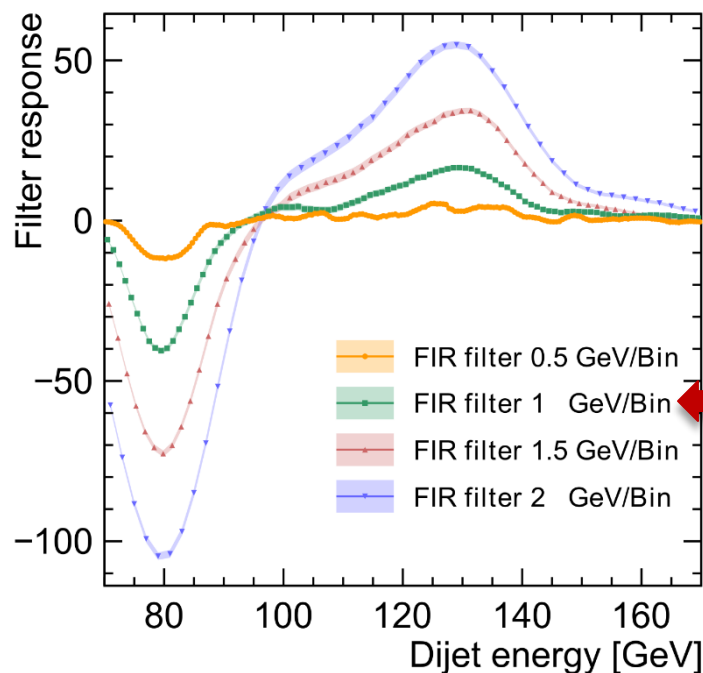
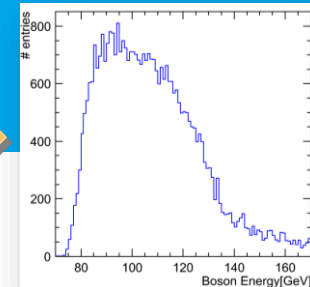
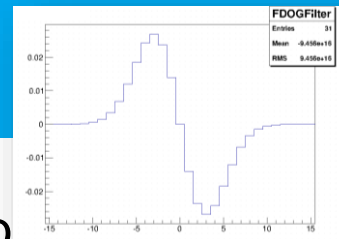
- The width of the Gaussian ( $\sigma$ ) = 4
- **The kernel size (# bins of the filter histogram)** *(the  $\sigma$  and bin sizes were fixed)*
- The binning of the input boson energy histogram





# FDOG Filter Optimisation

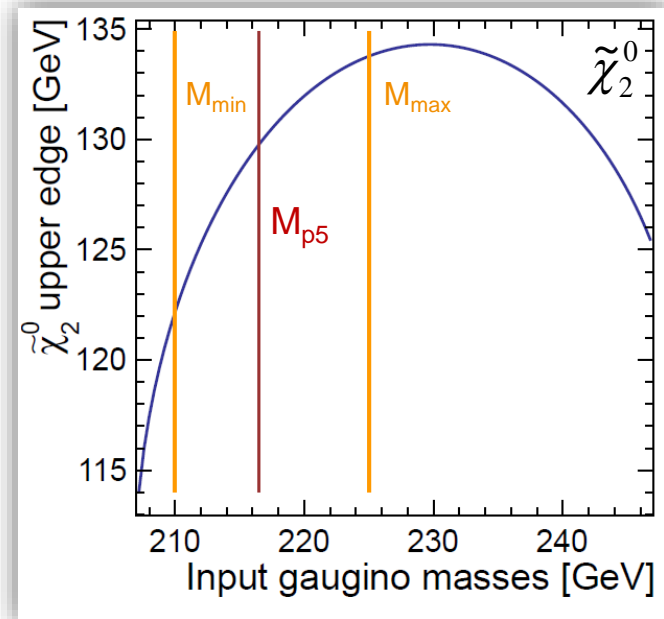
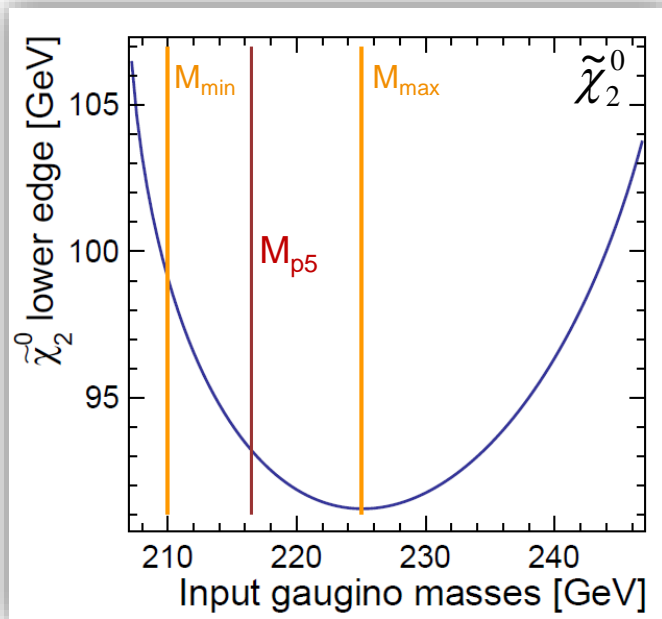
- There are 3 filter parameters that can be optimised
- The width of the Gaussian ( $\sigma$ ) = 4
  - The kernel size (# bins of the filter histogram) = 17
  - **The binning of the input boson energy histogram** (*the  $\sigma$  and kernel sizes were fixed*)



# Edge Calibration

- > The relation edge position  $\leftrightarrow$  input gaugino mass is given by:

$$E_V = \frac{M_{\tilde{\chi}}^2 + M_V^2 - M_{LSP}^2}{2M_{\tilde{\chi}}} \quad \text{and} \quad E_V^{lab} = \gamma E_V \pm \beta \gamma \sqrt{E_V^2 - M_V^2} \quad (\text{NO ISR, beamstrahlung...})$$



- > Ignore  $\tilde{\chi}_1^\pm$  low edge
- > Chosen mass range:  $M_{\tilde{\chi}}^{\min} = 210$  GeV  $\leftrightarrow$   $M_{\tilde{\chi}}^{\max} = 225$  GeV, in steps of 3 GeV
- > Generate the **same number of signal  $\tilde{\chi}_1^\pm$  and  $\tilde{\chi}_2^0$  events** for all samples
- > The **SM background** is the **same** for all mass samples

# Choosing the Appropriate Filter

## > Canny's criteria for an optimal filter:

- **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

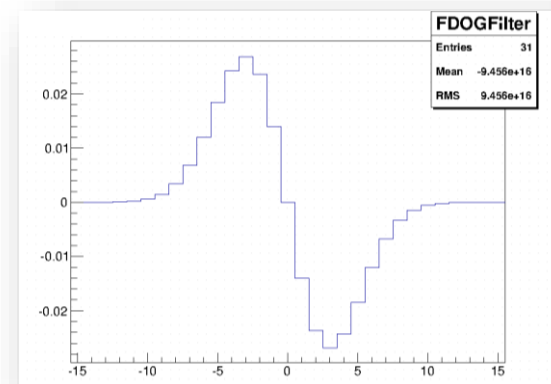
## > Canny has shown that an optimal filter is very similar to the first derivative of a Gaussian

## > There are 3 filter parameters that can be optimised (via toy Monte Carlo)

- The width of the Gaussian ( $\sigma$ ) = **4**
- The kernel size (# bins of the filter histogram) = **17**
- The binning of the input boson energy histogram = **1 GeV/bin**

## > Edge positions stable within max.1.8% when varying filter parameters

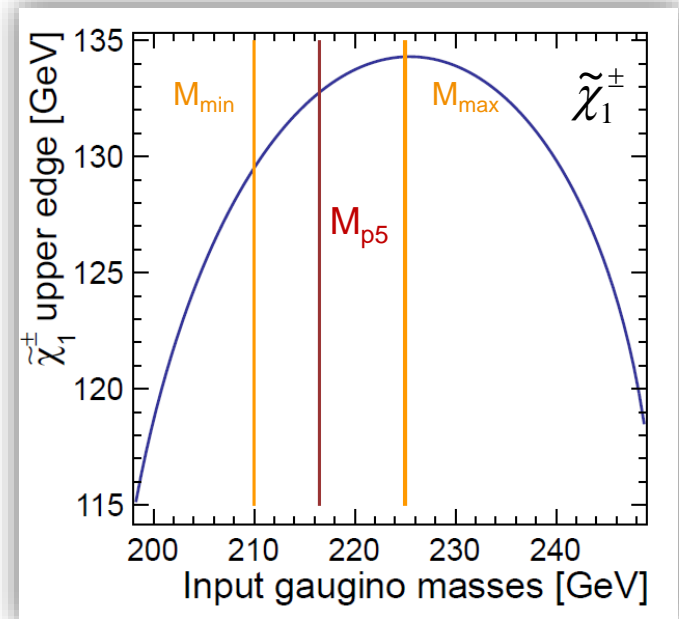
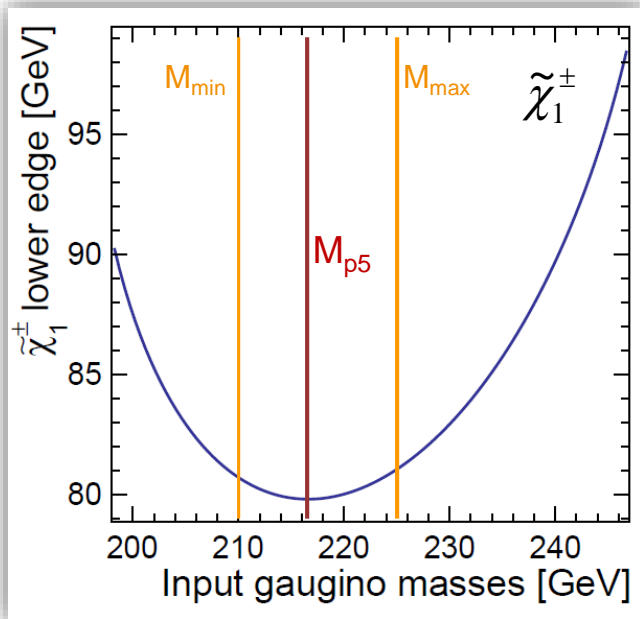
## > (Reminder: LOI edge fluctuations [from LOI vs DBD comparison]: 9.4%)



# Edge Calibration

- > The relation edge position  $\leftrightarrow$  input gaugino mass is given by:

$$E_V = \frac{M_\chi^2 + M_V^2 - M_{LSP}^2}{2M_\chi} \quad \text{and} \quad E_V^{lab} = \gamma E_V \pm \beta \gamma \sqrt{E_V^2 - M_V^2} \quad (\text{NO ISR, beamstrahlung...})$$

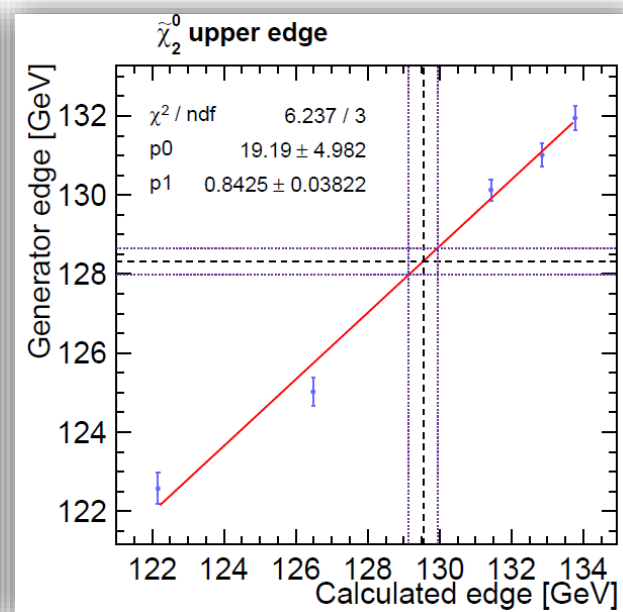
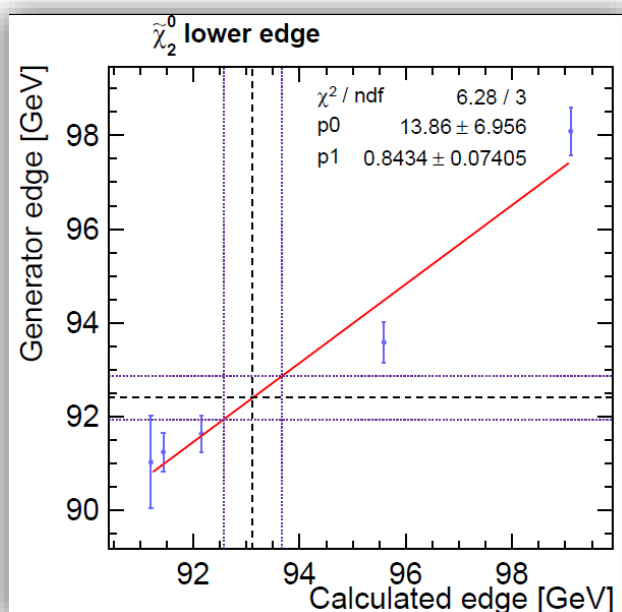
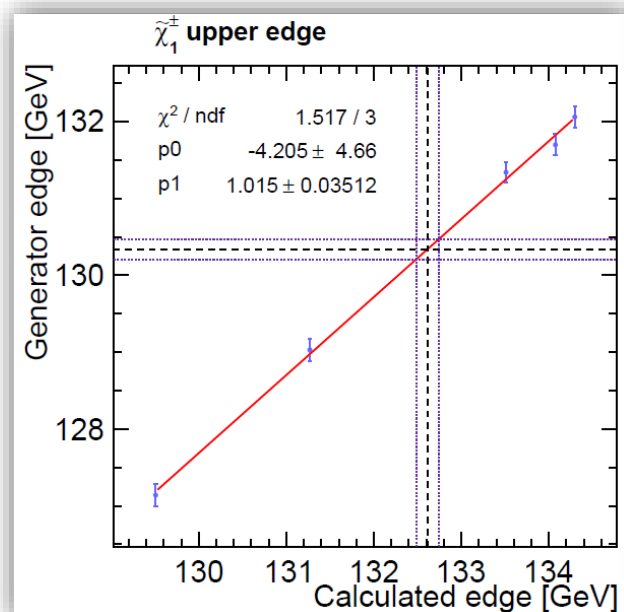


# Edge Calibration II

## ➤ Three different aspects:

1. Calibrate **edges** measured **on generator level** w.r.t. **calculated edges**

→ **study effects of ISR emission, beamstrahlung**



# Edge Calibration II

## ➤ Three different aspects:

1. Calibrate **edges** measured **on generator level** w.r.t. **calculated edges**

└→ **study effects of ISR emission, beamstrahlung**

Gaugino	Generator [GeV]		Calculated [GeV]		Calibrated [GeV]	
	$E_{low}$	$E_{high}$	$E_{low}$	$E_{high}$	$E_{low}$	$E_{high}$
$\tilde{\chi}_1^\pm$	—	$130.34 \pm 0.13$	—	132.76	—	$132.61 \pm 0.13$
$\tilde{\chi}_2^0$	$92.395 \pm 0.46$	$128.32 \pm 0.34$	93.09	129.92	$93.11 \pm 0.55$	$129.54 \pm 0.4$

- Beam effects have an impact of 0.8% → 1.8%



# Edge Calibration II

## ➤ Three different aspects:

1. Calibrate **edges** measured **on generator level** w.r.t. **calculated edges**  
└→ **study effects of ISR emission, beamstrahlung [0.8% → 1.8%]**
2. Calibrate **edges** measured **on reconstruction level** w.r.t. **generator level edges**  
└→ **study simulation and reconstruction effects**

Gaugino	Reconstructed [GeV]		Generator [GeV]		Calibrated [GeV]	
	$E_{low}$	$E_{high}$	$E_{low}$	$E_{high}$	$E_{low}$	$E_{high}$
$\tilde{\chi}_1^\pm$	—	$129.21 \pm 0.299$	—	$130.34 \pm 0.13$	—	$130.39 \pm 0.28$
$\tilde{\chi}_2^0$	$92.62 \pm 0.24$	$127.82 \pm 0.35$	$92.395 \pm 0.46$	$128.32 \pm 0.34$	$92.21 \pm 0.29$	$128.64 \pm 0.31$

- **Simulation and reconstruction effects have an impact of 0.2% → 0.9% !**



# Edge Calibration Results II

## ➤ Three different aspects:

1. Calibrate **edges** measured **on generator level** w.r.t. **calculated edges**  
→ study effects of ISR emission, beamstrahlung [0.8% → 1.8%]
2. Calibrate **edges** measured **on reconstruction level** w.r.t. **generator level edges**  
→ study simulation and reconstruction effects [0.2% → 0.9%]
3. Calibrate **edges** measured **on reconstruction level** w.r.t. **calculated edges**  
→ take all the effects into account [1.1 → 2%]

Gaugino	Reconstructed [GeV]		Calculated [GeV]		Calibrated [GeV]	
	$E_{low}$	$E_{high}$	$E_{low}$	$E_{high}$	$E_{low}$	$E_{high}$
$\tilde{\chi}_1^\pm$	—	$130.13 \pm 0.78$	—	$132.77$	—	$132.79 \pm 0.44$
$\tilde{\chi}_2^0$	$92.11 \pm 0.31$	$128.99 \pm 0.75$	$93.09$	$129.92$	$92.52 \pm 1.23$	$130.67 \pm 0.77$

- Cumulative effects have an impact of 1.1% → 2% !