

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

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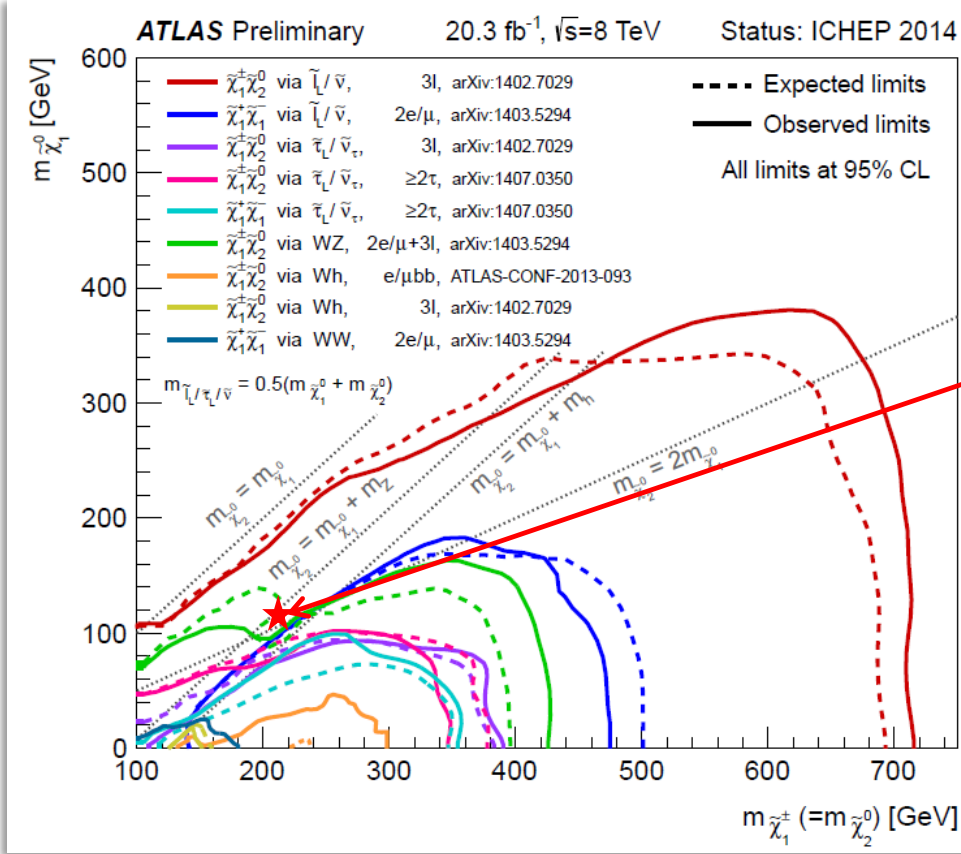


Study case: $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_2^0$ Pair Production at the ILC

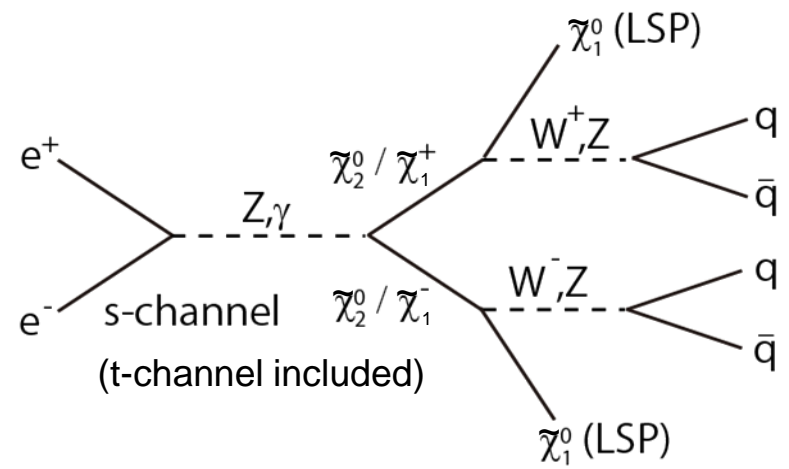
“Point 5” benchmark : gaugino pair production at ILC

<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 $\gamma\gamma \rightarrow$ 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%
Selection for x-section	Efficiency	72%	73%
	Purity	27%	5%



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: $(E, p \text{ conservation})$

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

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:

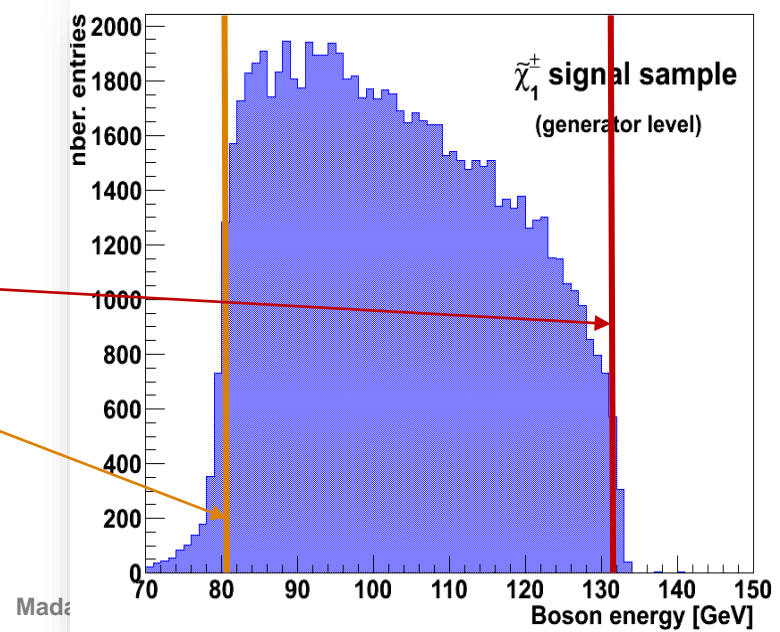
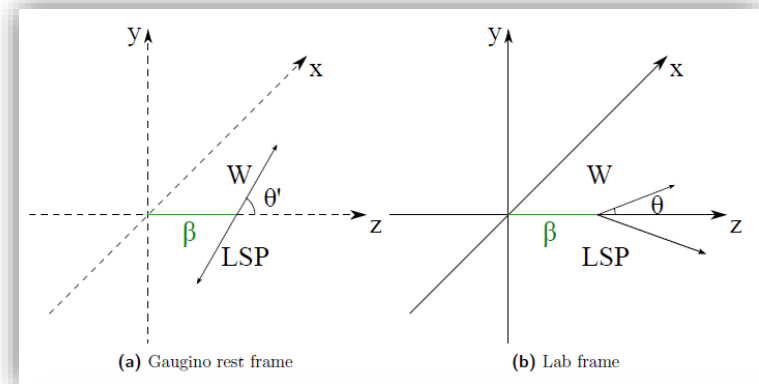
$$E_V^{lab} = \gamma E_V + \beta \gamma \vec{p}_{V,\parallel}$$

$$= \gamma E_V + \beta \gamma |\vec{p}_V| \cos \theta'$$

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

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

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



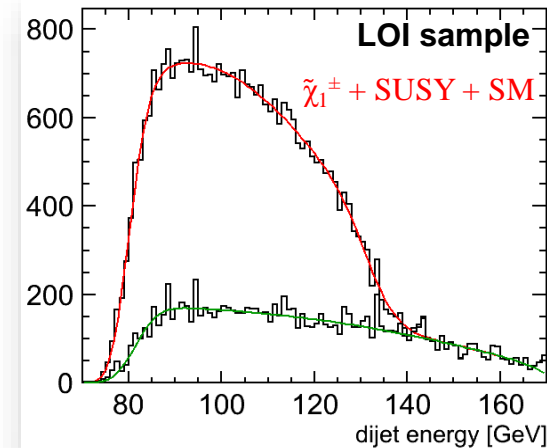
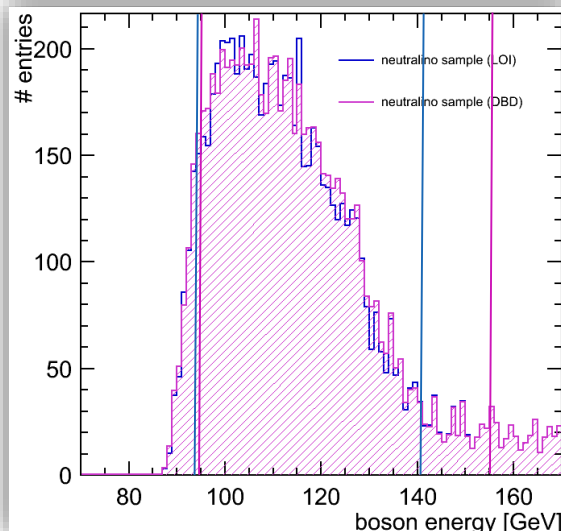
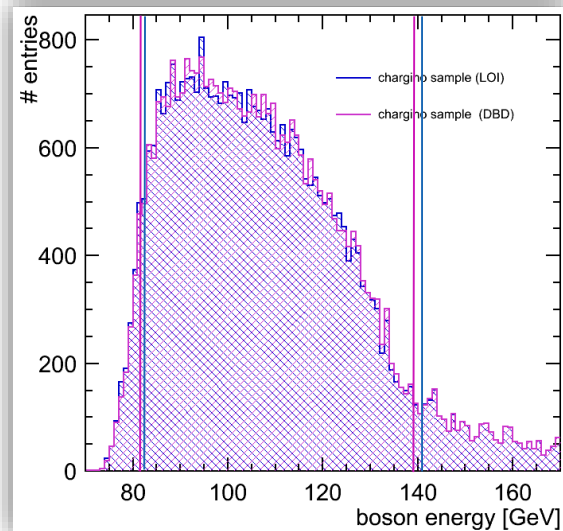
LOI Strategy: Fit the Boson Energy Spectrum

- Fit dijet energy spectrum and obtain edge positions:

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

- 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.

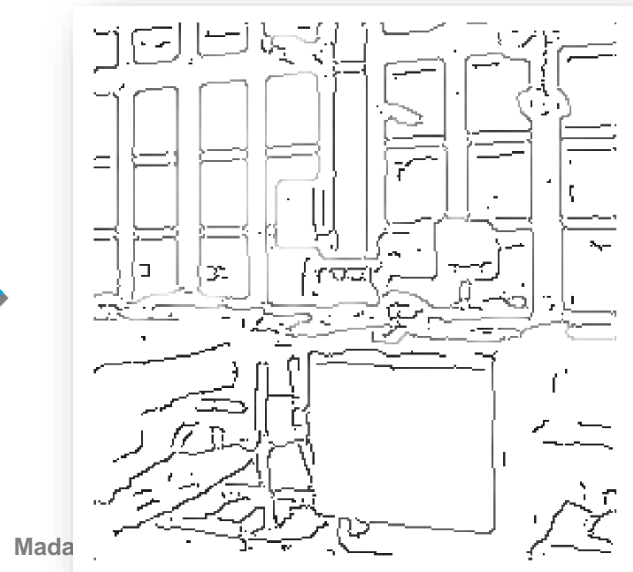
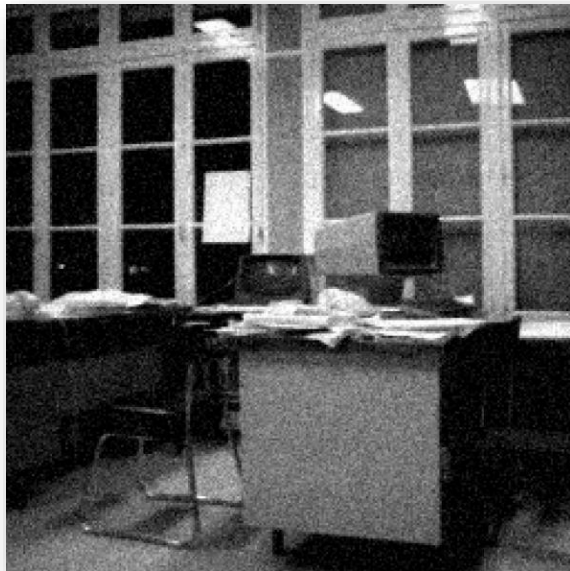
$$y[n] = \sum_{k=-M_1}^{M_2} b_k x[n - k]$$

the input signal

the filter coefficients (weights)

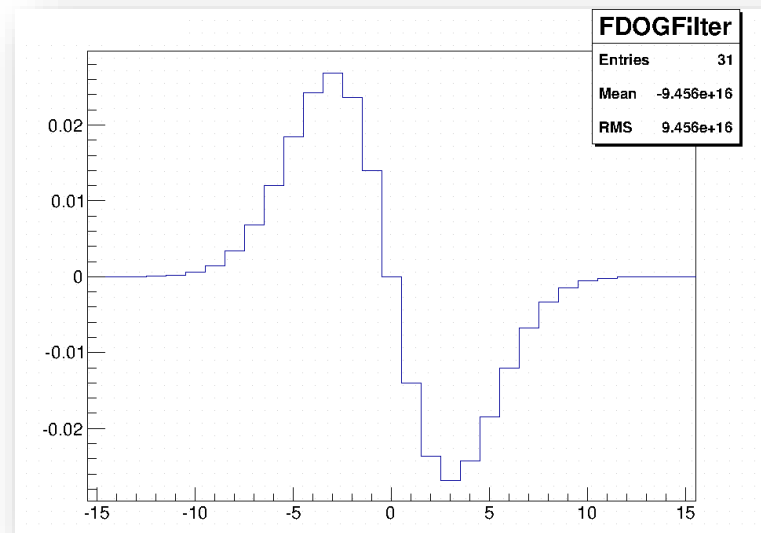
- y is obtained by convolving the input signal with the (finite) weights

D. Demigny, T. Kamlé

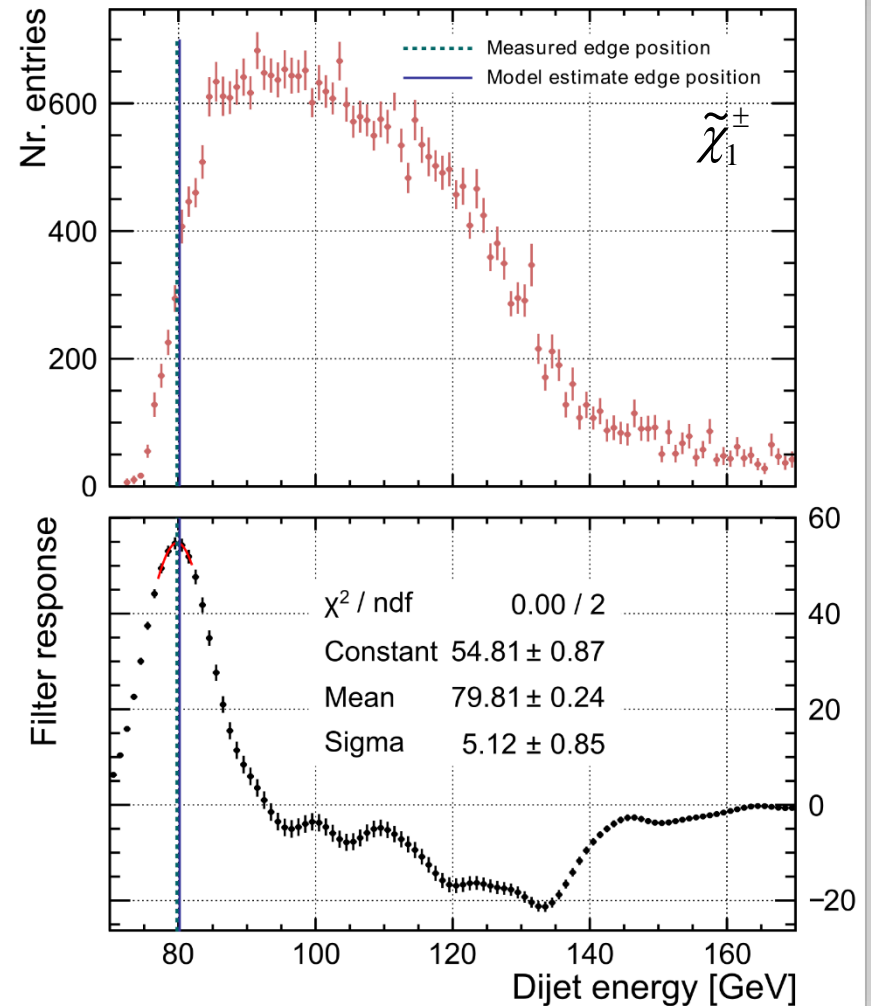
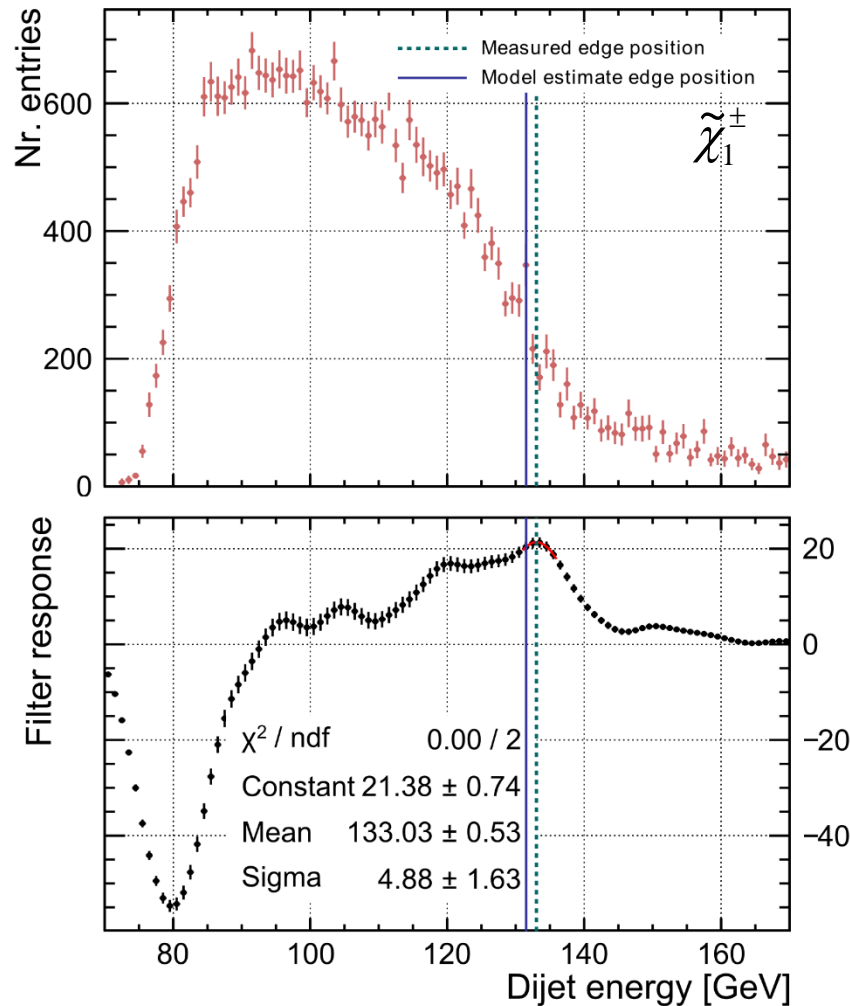
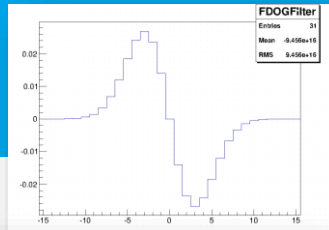


Choosing the Appropriate Filter

- > 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_{low} [GeV]	Edge W_{high} [GeV]	Edge Z_{low} [GeV]	Edge Z_{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

filter

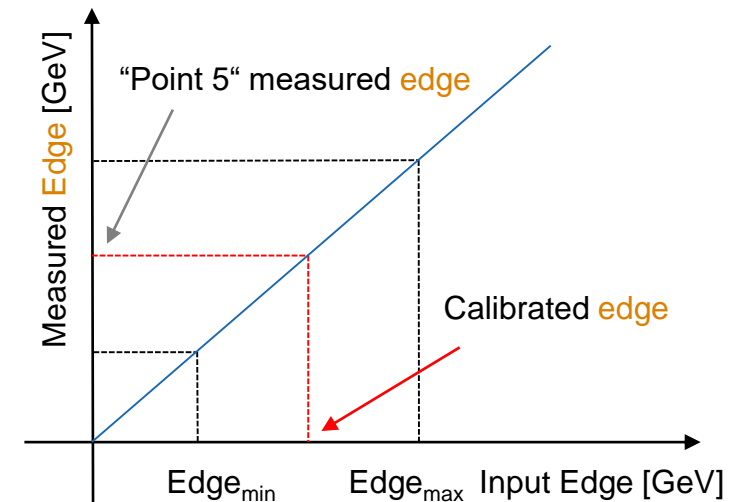
Sample	Mass $\tilde{\chi}_1^\pm$ [GeV]	Mass $\tilde{\chi}_2^0$ [GeV]	Mass $\tilde{\chi}_1^0$ [GeV]
TRUE	216.5	216.7	115.7
LOI	216.9±3.2	220.0±1.4	118.4±1.1
DBD	216.8±3.2	220.6±1.2	118.2±0.9

- 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

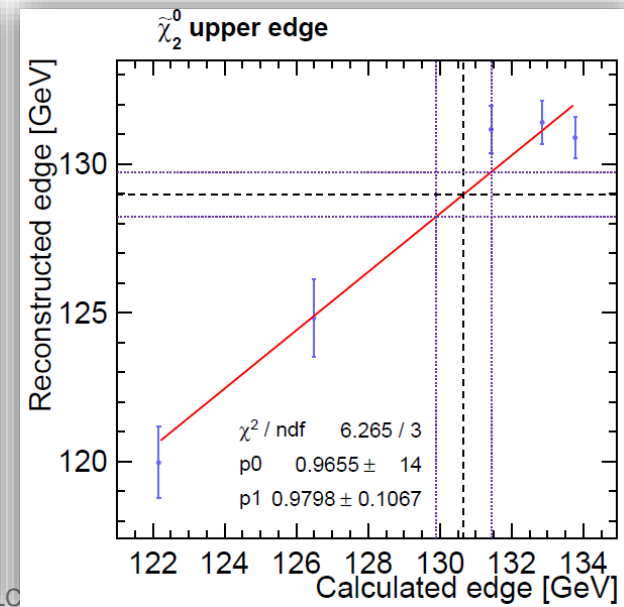
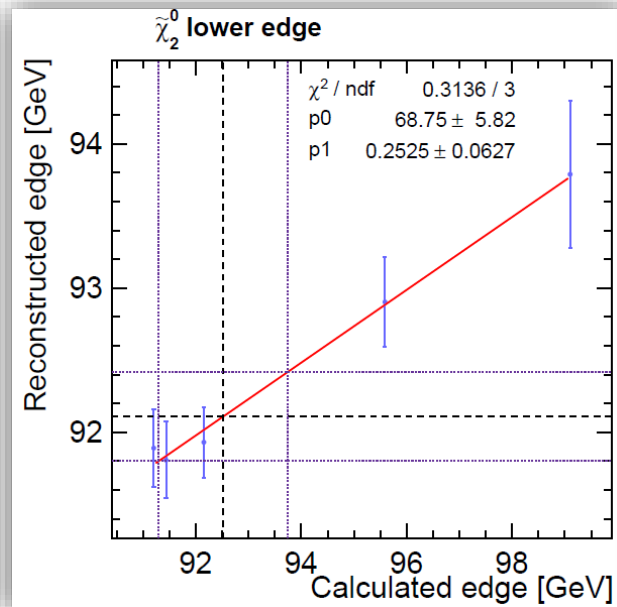
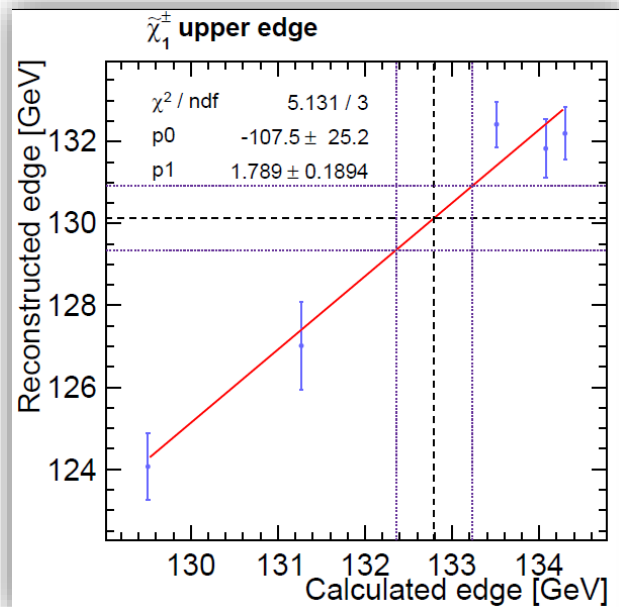
- Performed **only** for **DBD** sample → account for systematics
- Calibrate the edge positions → then calculate the calibrated mass(es)
- Edge calibration procedure:
 - Vary input masses: $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_2^0$ varied **simultaneously**, **LSP** mass **fixed!**
 $M_\chi^{min}=210$ 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 ± 3.1	214.1 ± 4.8	220.9 ± 2.9	216.5 [GeV]
$\tilde{\chi}_2^0$	220.4 ± 1.3	216.9 ± 3.4	220.6 ± 1.7	216.7 [GeV]
$\tilde{\chi}_1^0$	118.1 ± 0.9	115.5 ± 1.8	118.9 ± 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

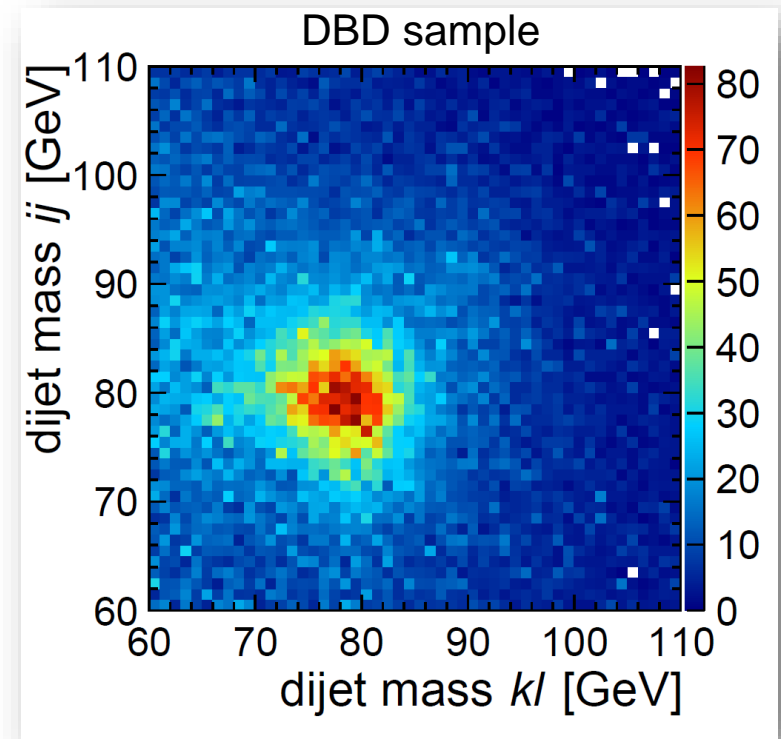
> Relevant distribution: the reconstructed mass of one dijet pair versus the other:

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

> Since $\sigma \propto \frac{Nr.\text{events}}{\varepsilon \cdot \int \mathcal{L}} \Rightarrow$ 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.



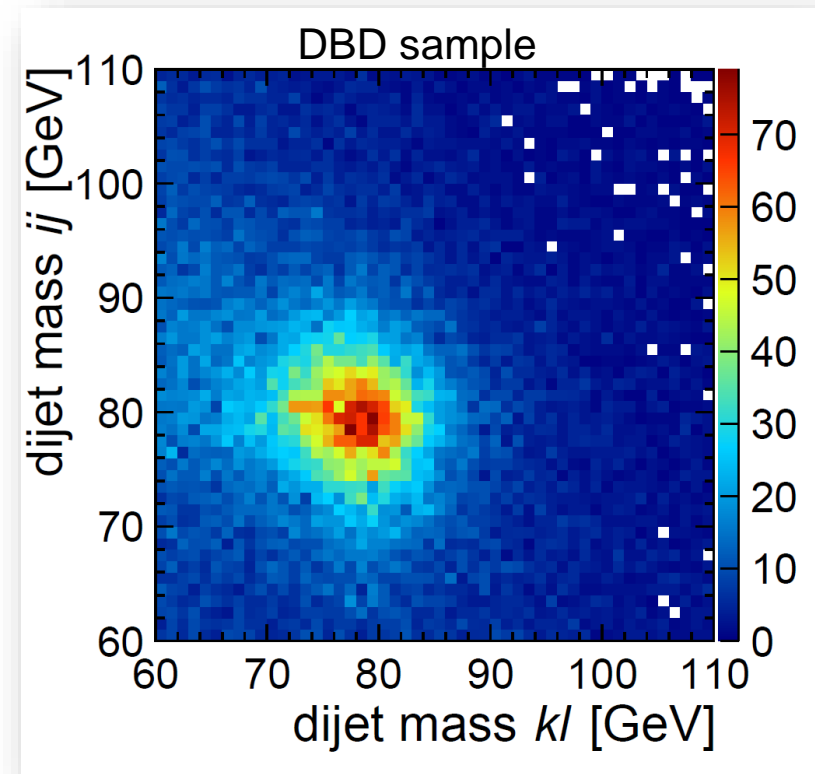
The total distribution (SUSY + SM)



Cross Section: 2D Template Fit

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

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



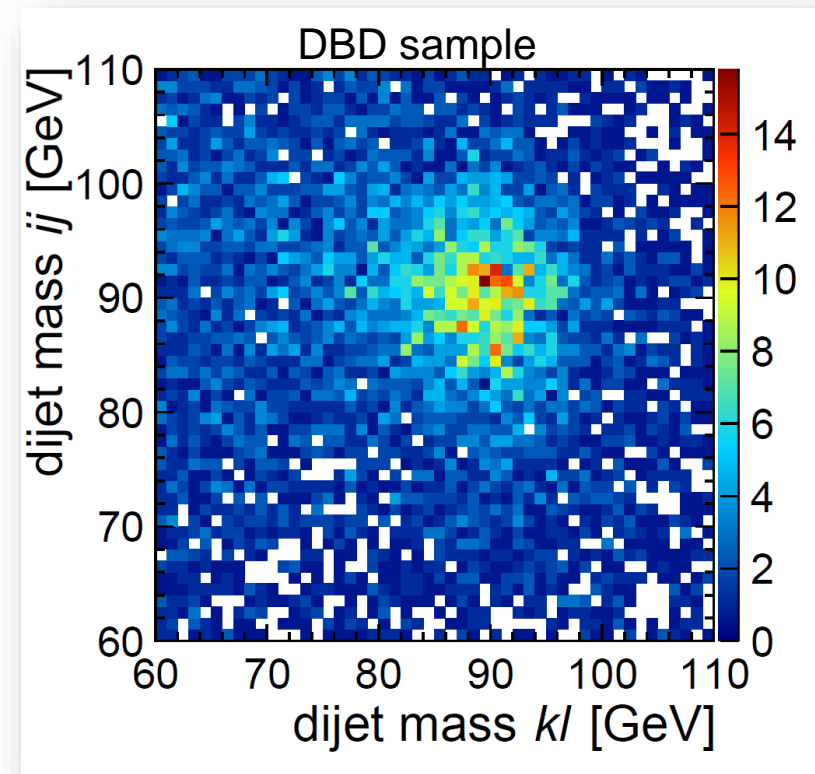
Chargino events only

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



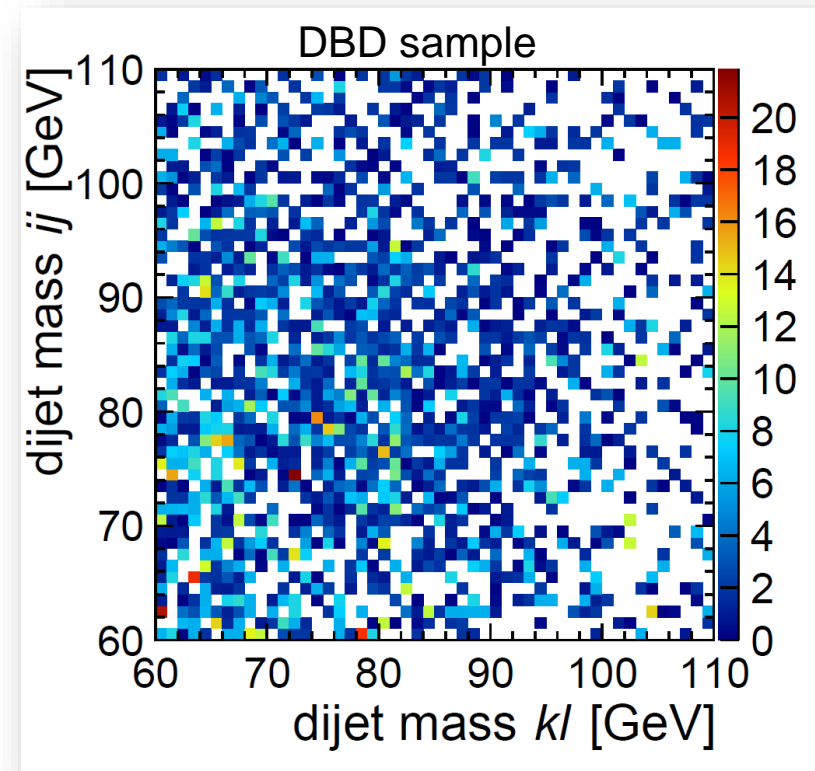
Neutralino events only

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

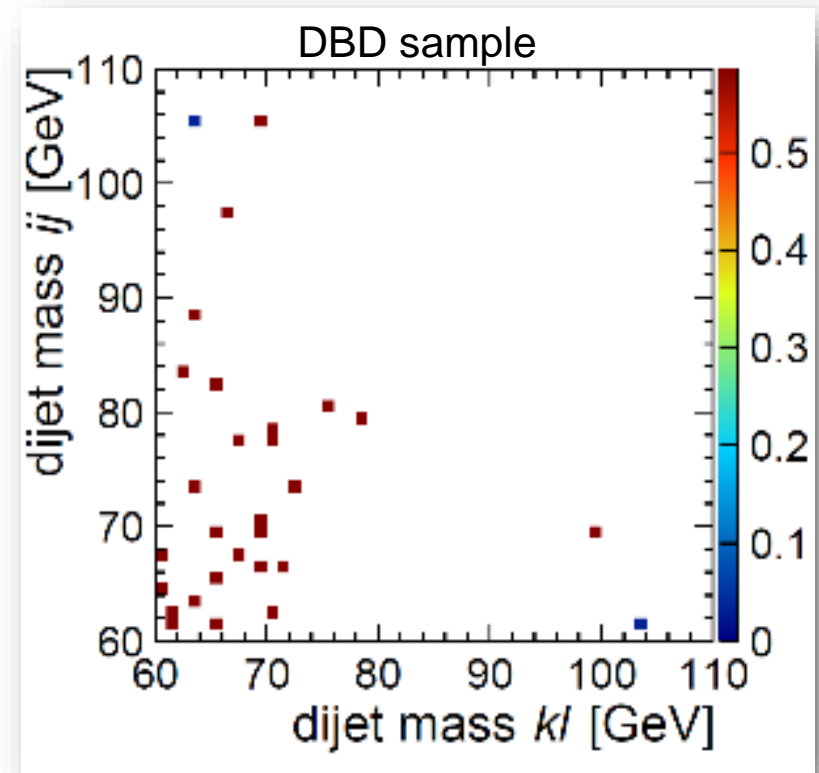


Standard Model events only

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!**

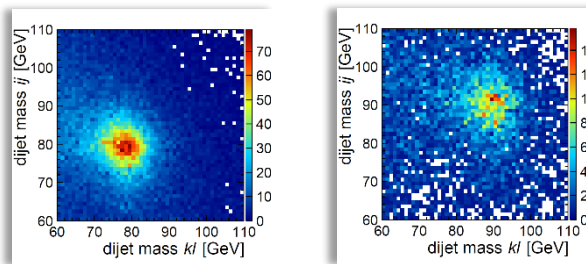
- After preselection
- Kinematic fit converged
- All dijet permutations included



SUSY background events only

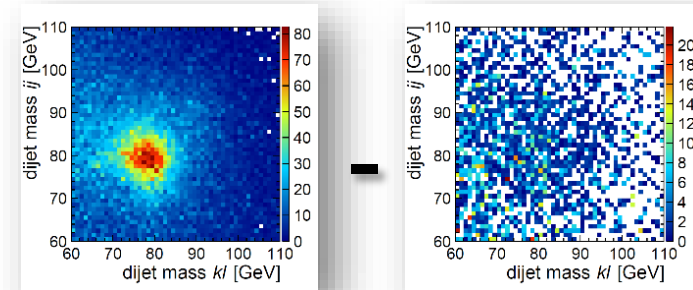
Cross Section: 2D Template Fit

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



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

- a and b → 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

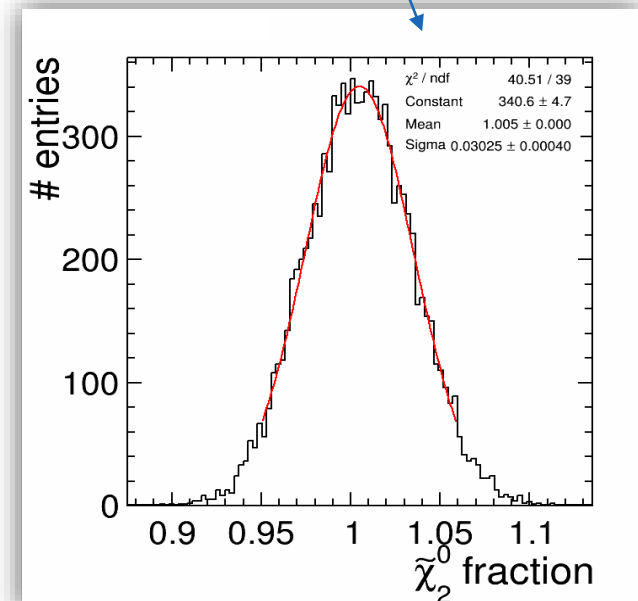
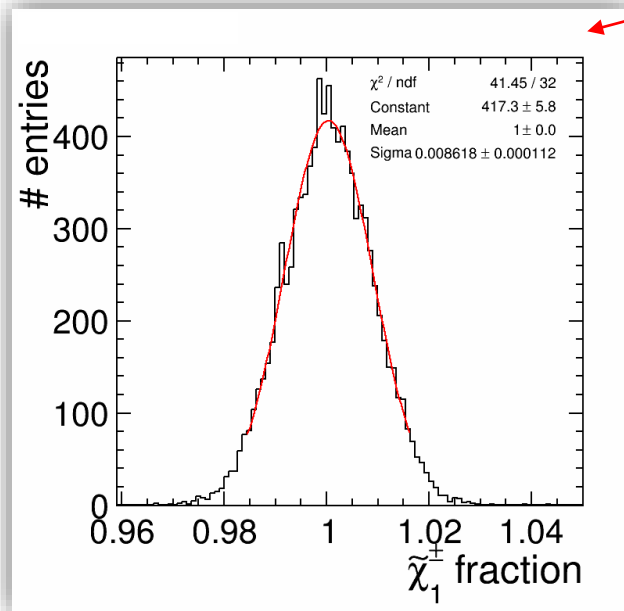


2D Template Fit Toy Monte Carlo

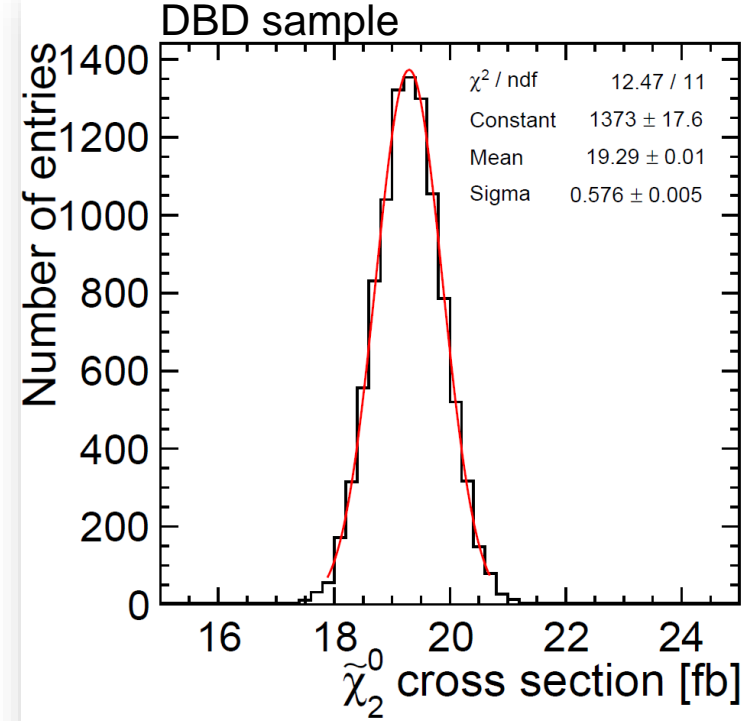
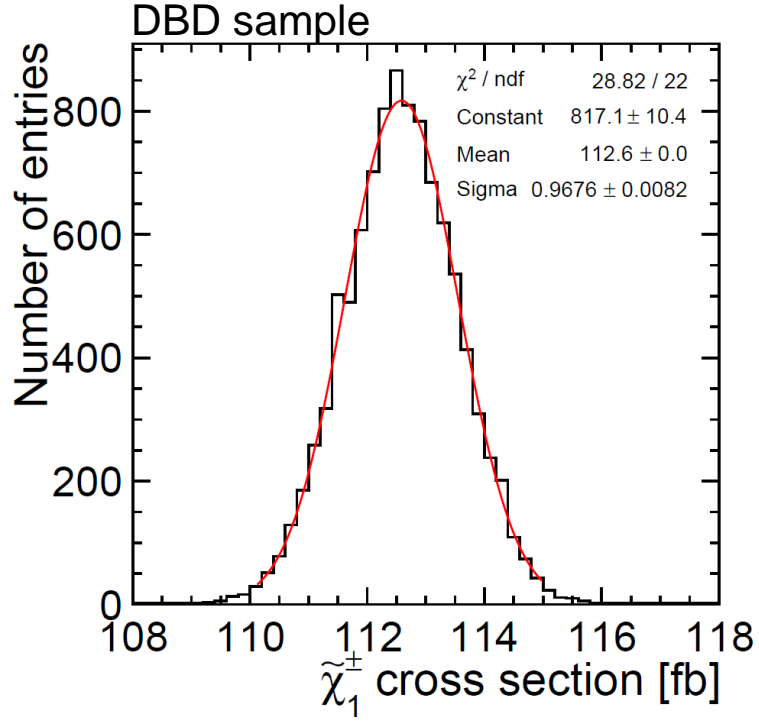
> **Note:** limited amount of Monte Carlo data available → **toy Monte Carlo study**

> **Running the toy MC:**

- Treat the total data distribution as a p.d.f
- Randomly sample the initial distribution N times: $N = N_{evts.}^{initial} \pm \sqrt{N_{evts.}^{initial}}$
- 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 ± 0.97	19.3 ± 0.58



Cross Section: 2D Template Fit – Comparison to LOI

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

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

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

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

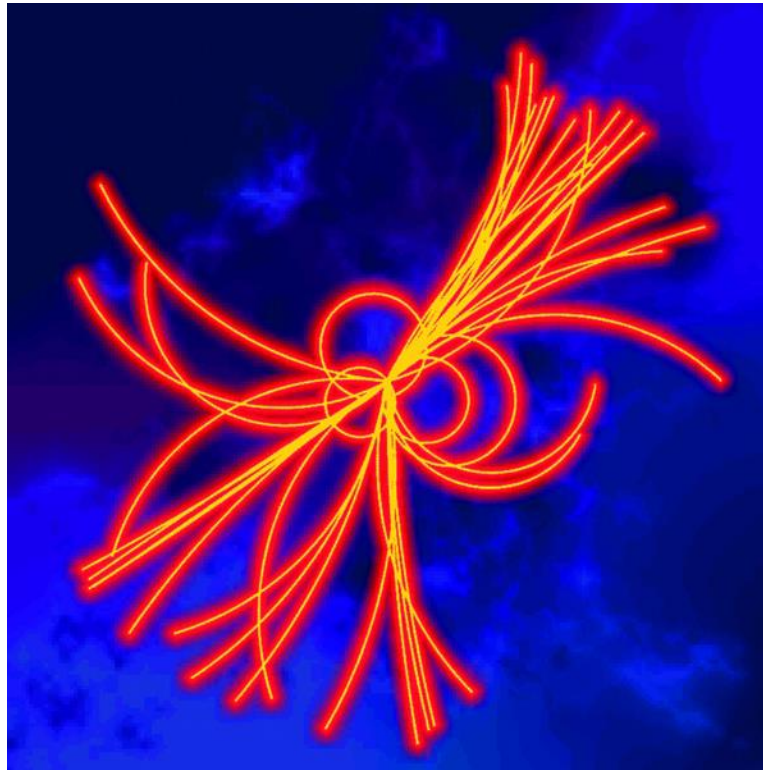


Conclusions

- 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

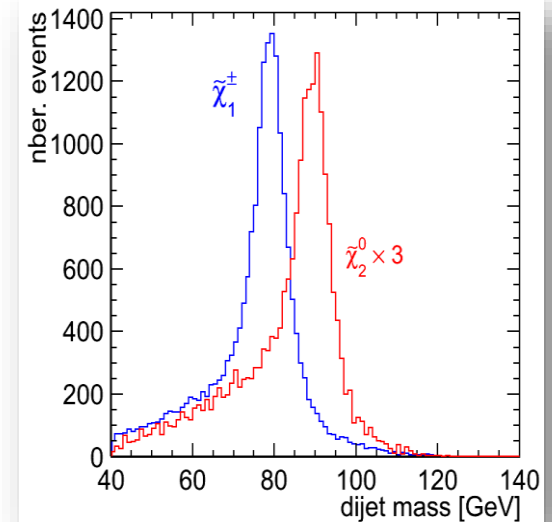
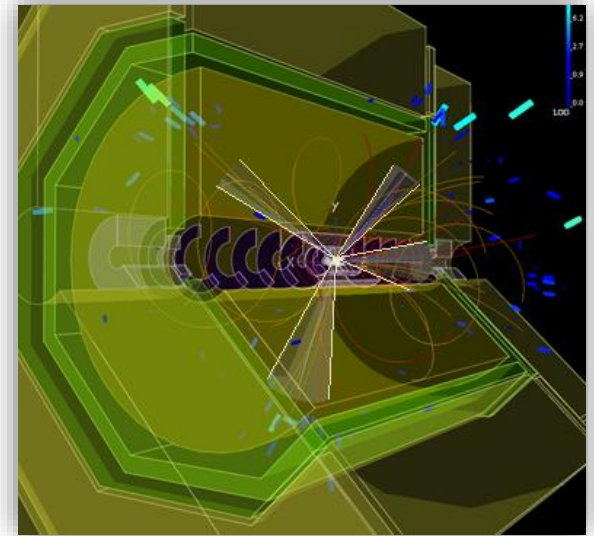
➤ Signal topology:

- **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

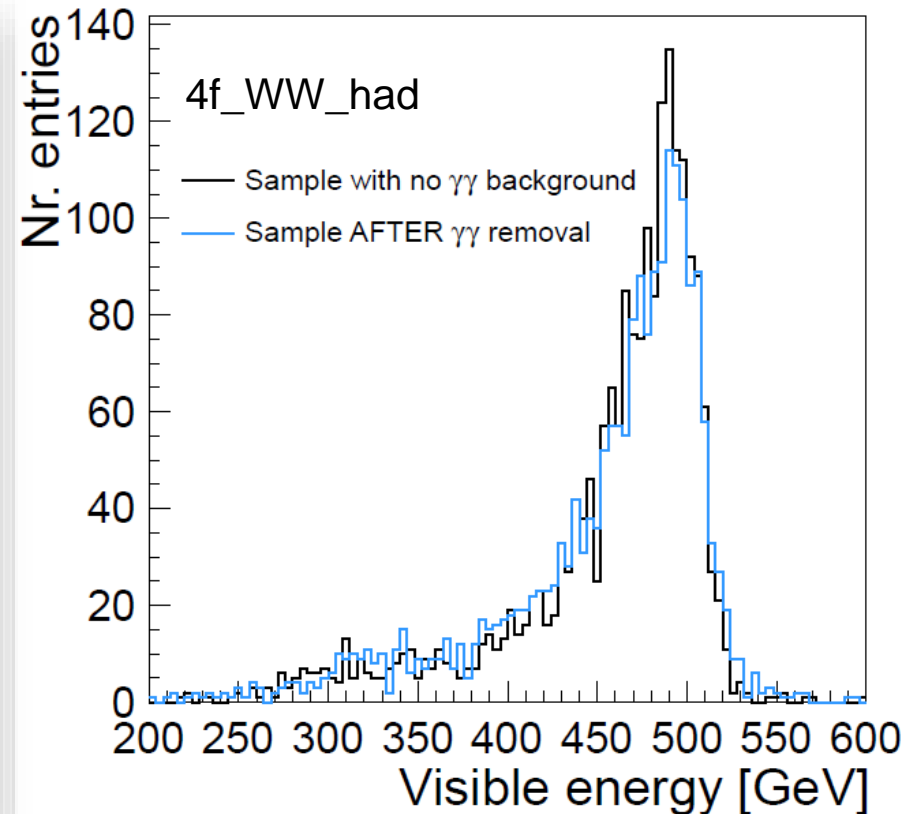
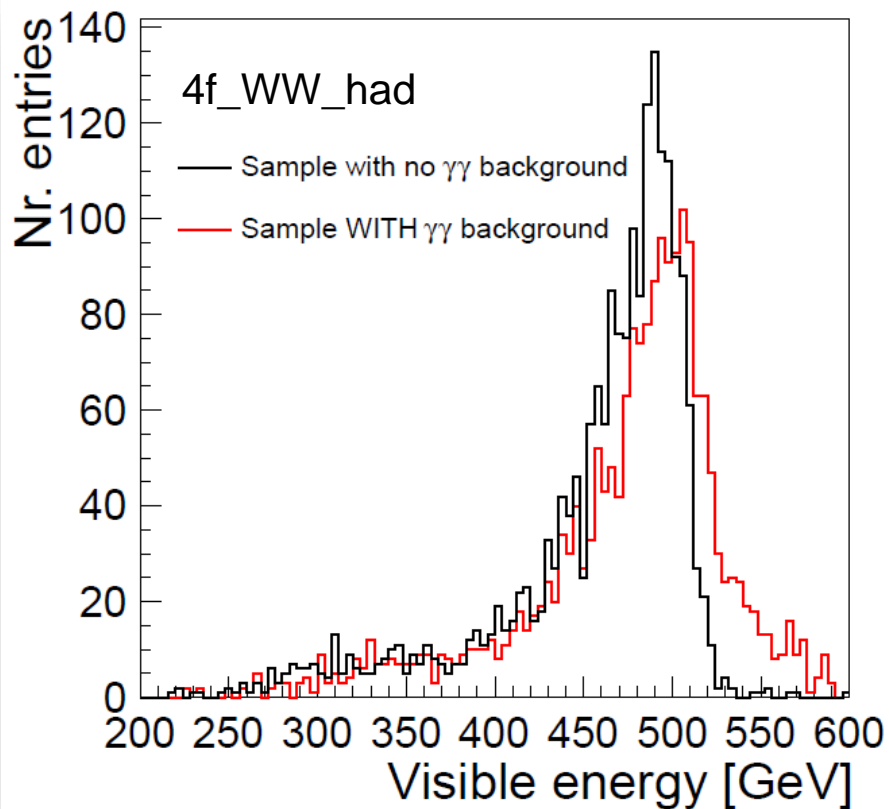
- Signal + background were simulated and reconstructed with `ilcsoft v01-06`
- The jet energy scale was increased by 1%
- No $\gamma\gamma$ background overlay
- The analysis was run on existing data samples

- 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 $\gamma\gamma \rightarrow$ 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	$\gamma\gamma$ & γe (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

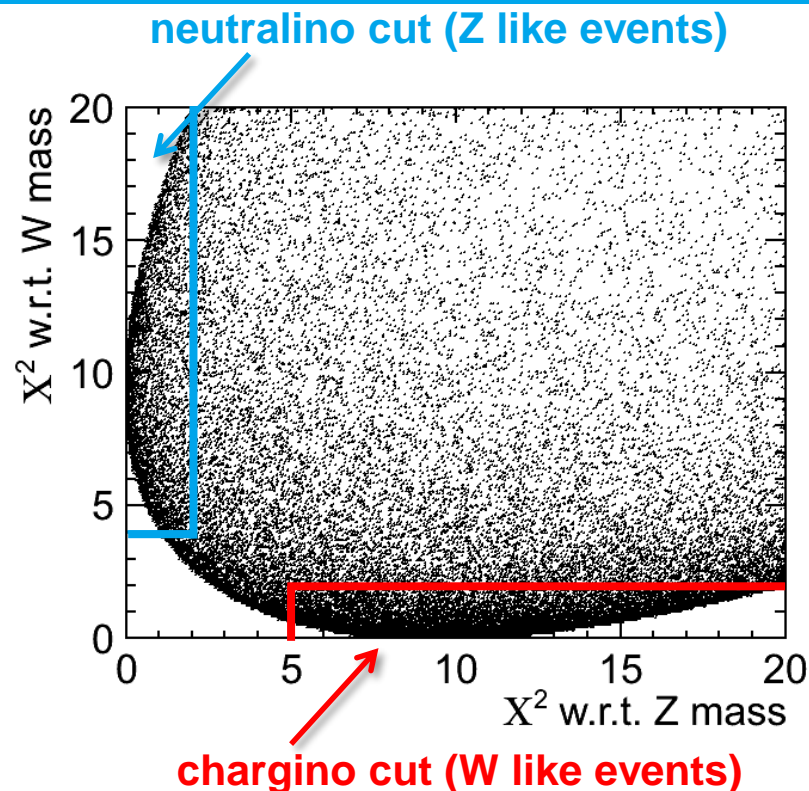
- > Calculate χ^2 with respect to nominal W / Z mass

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



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 χ^2 cut, the separation is quite clear:



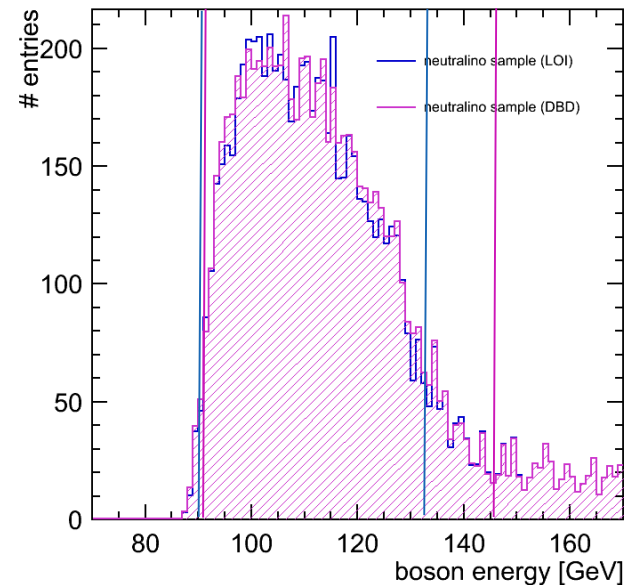
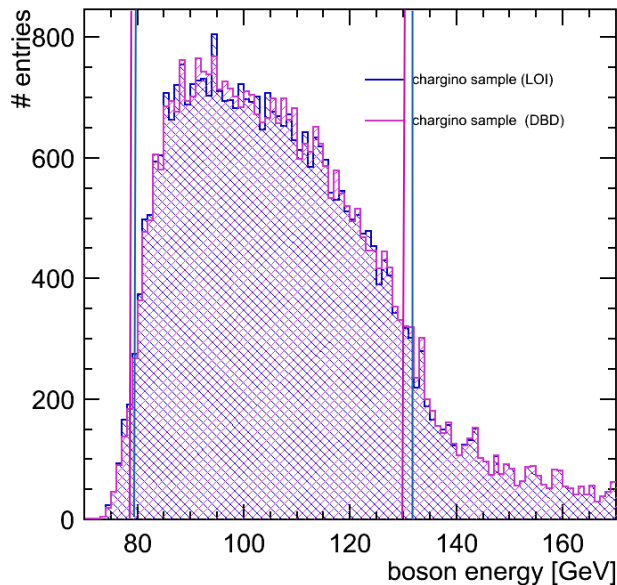
Sample	$\tilde{\chi}_1^\pm$ hadronic	$\tilde{\chi}_2^0$ hadronic
Efficiency	90.8%	91%
Purity	14.7%	2.6%



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

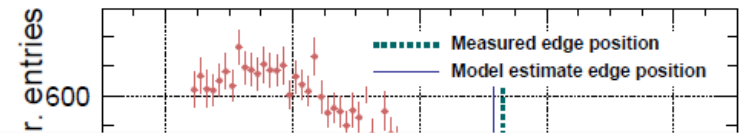
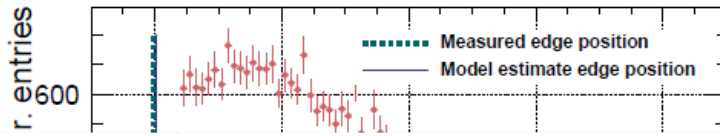


Sim.	Edge W_{low} [GeV]	Edge W_{high} [GeV]	Edge Z_{low} [GeV]	Edge Z_{high} [GeV]
DBD	79.5 ± 0.5	130.2 ± 1.1	91.3 ± 0.6	146.1 ± 4.8
LOI	79.7 ± 0.3	131.9 ± 0.9	91.0 ± 0.7	133.6 ± 0.5

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

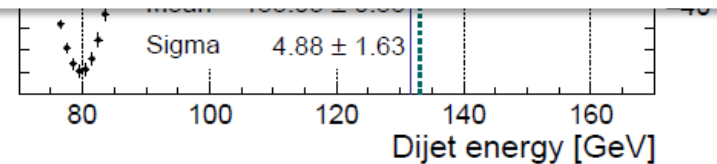
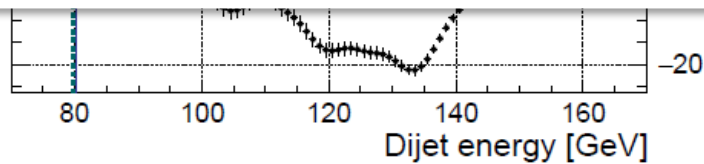
We need to apply a different edge extraction method!

Applying the FIR Filter on DBD Data: Results



filter

Calc.	80.17	131.53	93.24	129.06	
Sim.	Edge W_{low} [GeV]	Edge W_{high} [GeV]	Edge Z_{low} [GeV]	Edge Z_{high} [GeV]	
}	LOI	79.7 ± 0.3	131.9 ± 0.9	91.0 ± 0.7	133.6 ± 0.5
	DBD	79.5 ± 0.5	130.2 ± 1.1	91.3 ± 0.6	146.1 ± 4.8
	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

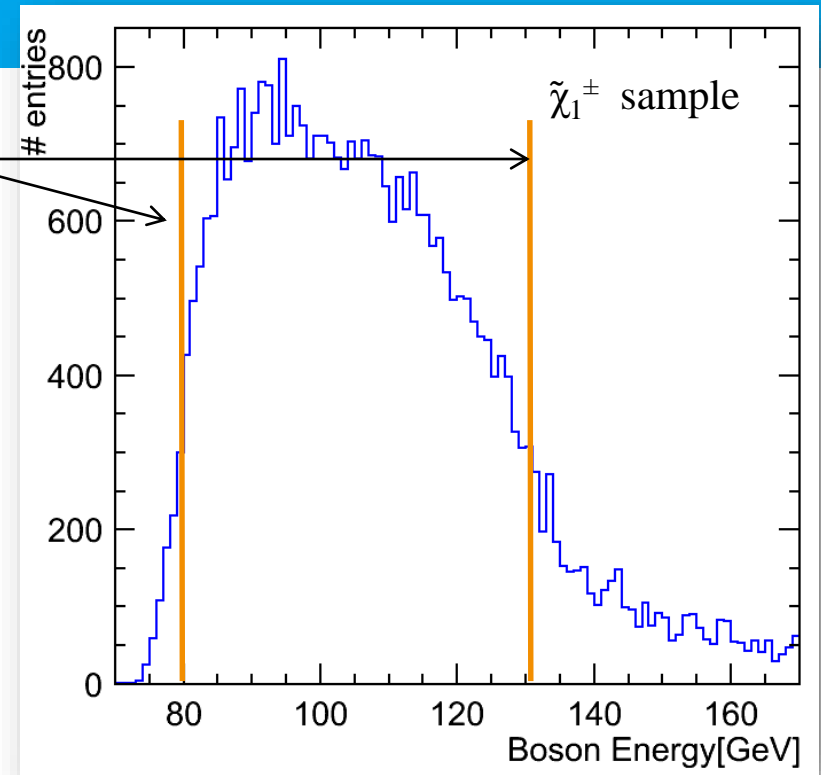


- **Statistical errors determined from toy Monte Carlo**



Applying an FIR Filter

- > Goal: find edge positions in spectrum



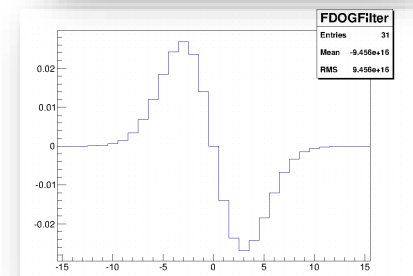
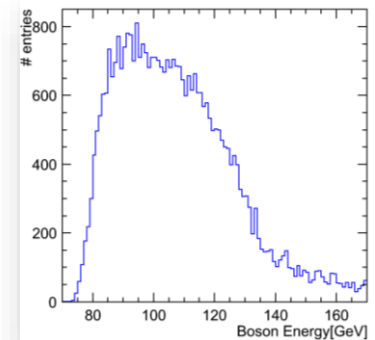
Applying an FIR Filter

- > 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
- 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

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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
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Bin #	1	2	3	...	98	99	100
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$$0 \times 15 + 0.01 \times 28 + \dots = \text{val2}$$

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



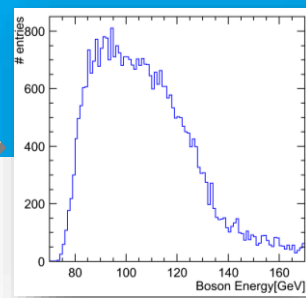
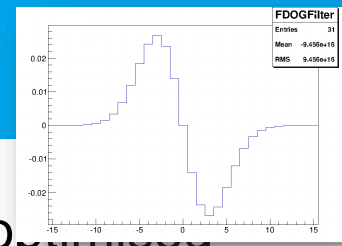
- There are 3 filter parameters that can be optimised
 - The width of the Gaussian (σ)
 - The kernel size (# bins of the filter histogram)
 - The binning of the input boson energy histogram



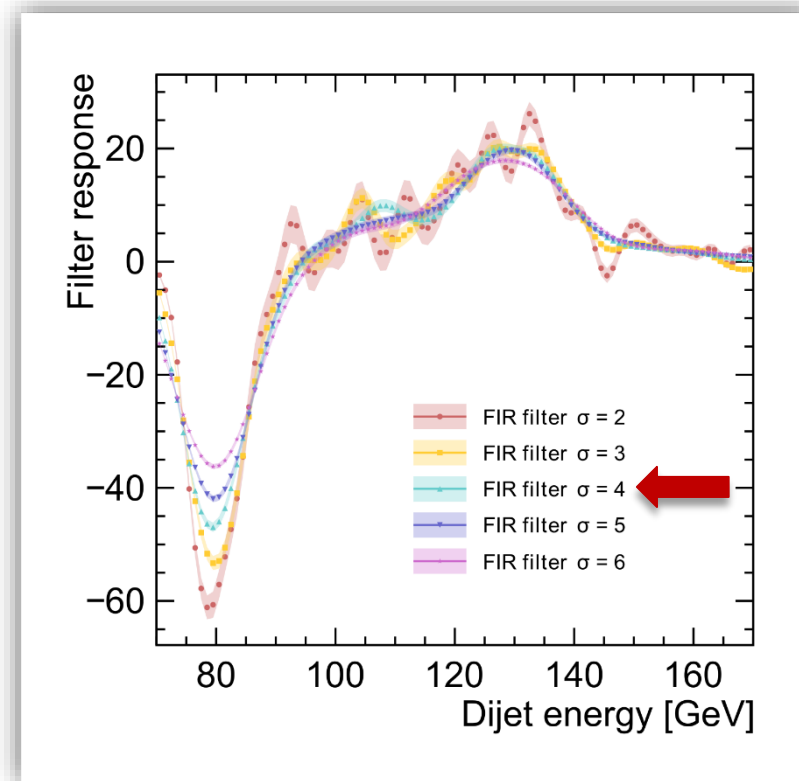
FDOG Filter Optimisation

> There are 3 filter parameters that can be optimised

- **The width of the Gaussian (σ)**
- The kernel size (# bins of the filter histogram)
- The binning of the input boson energy histogram



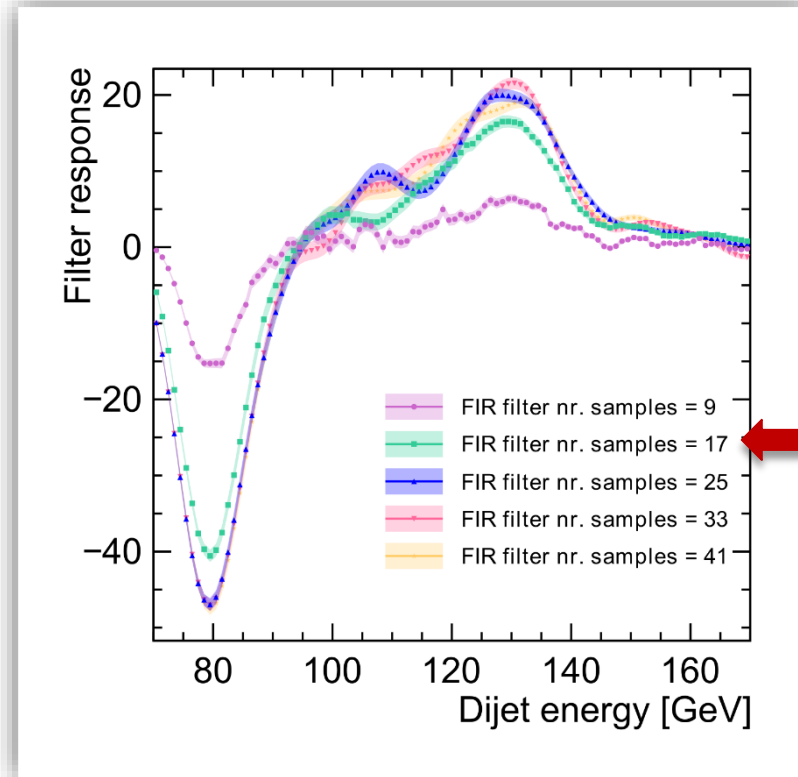
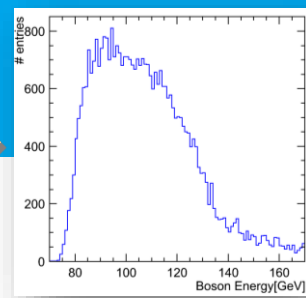
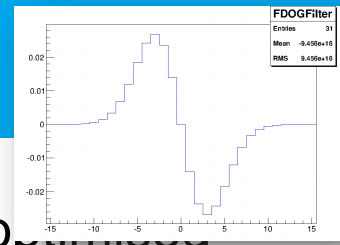
(the kernel and bin sizes were fixed)



FDOG Filter Optimisation

> There are 3 filter parameters that can be optimised

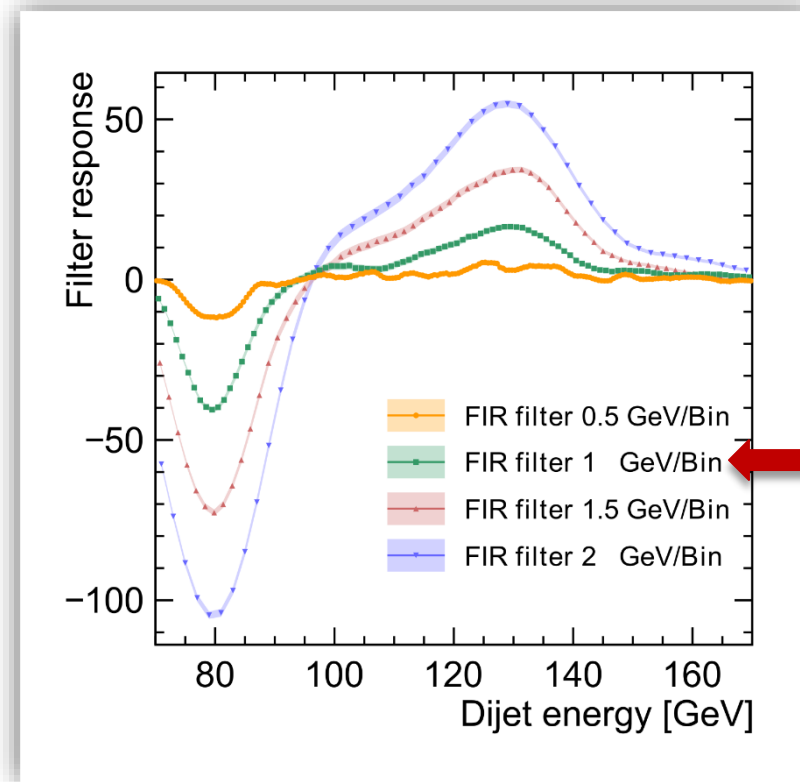
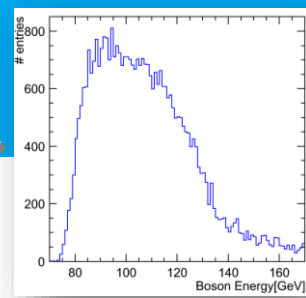
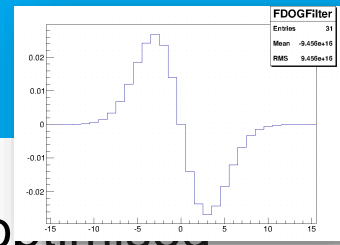
- The width of the Gaussian (σ) = 4
- **The kernel size (# bins of the filter histogram)** *(the σ and bin sizes were fixed)*
- The binning of the input boson energy histogram



FDOG Filter Optimisation

> There are 3 filter parameters that can be optimised

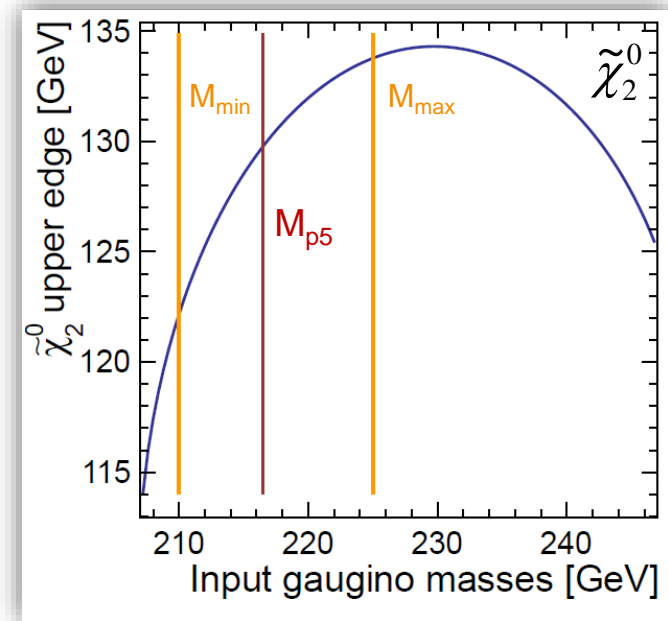
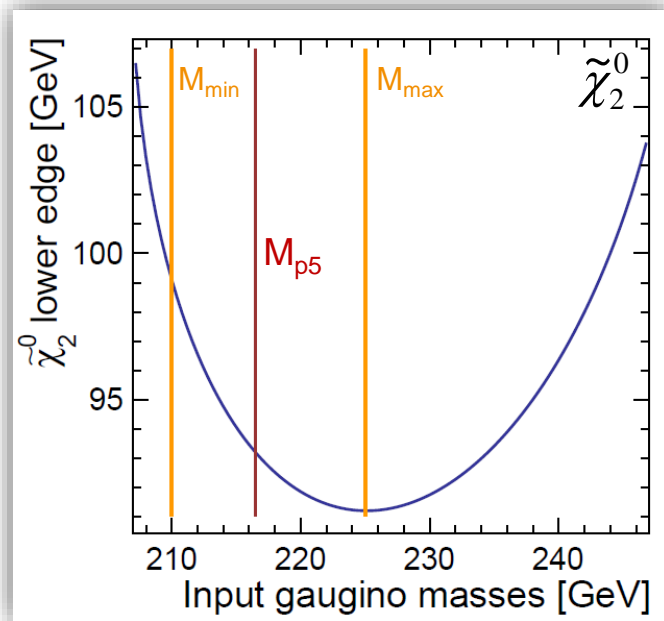
- The width of the Gaussian (σ) = 4
- The kernel size (# bins of the filter histogram) = 17
- **The binning of the input boson energy histogram** (*the σ and kernel sizes were fixed*)



Edge Calibration

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

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

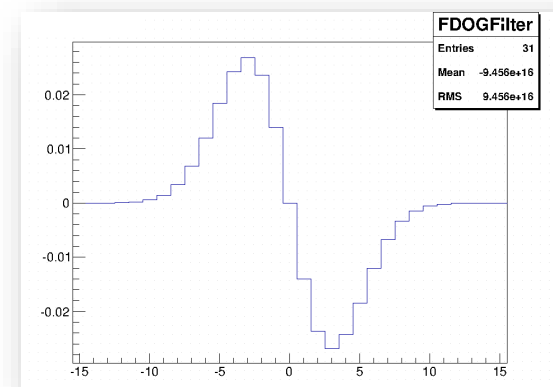


- > Ignore $\tilde{\chi}_1^\pm$ low edge
- > Chosen mass range: $M_{\tilde{\chi}}^{min}=210$ GeV \leftrightarrow $M_{\tilde{\chi}}^{max}=225$ GeV, in steps of 3 GeV
- > Generate the **same number of signal $\tilde{\chi}_1^\pm$ and $\tilde{\chi}_2^0$ events** for all samples
- > The **SM background**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



Choosing the Appropriate Filter

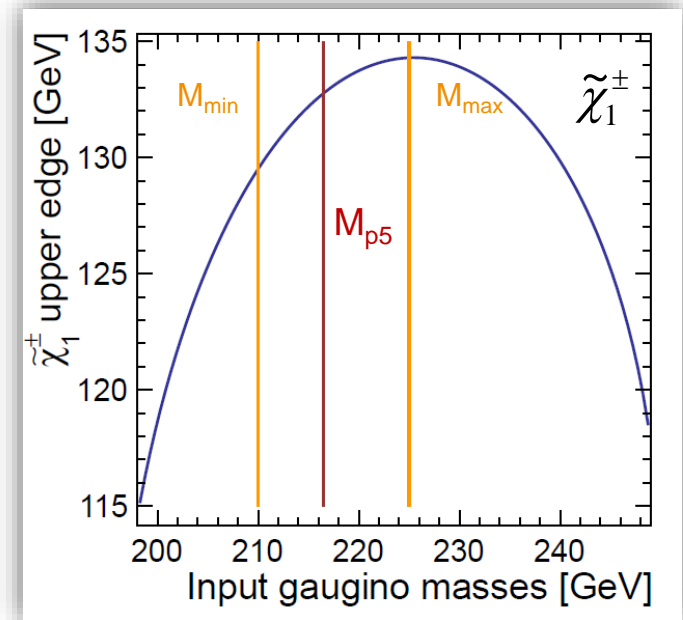
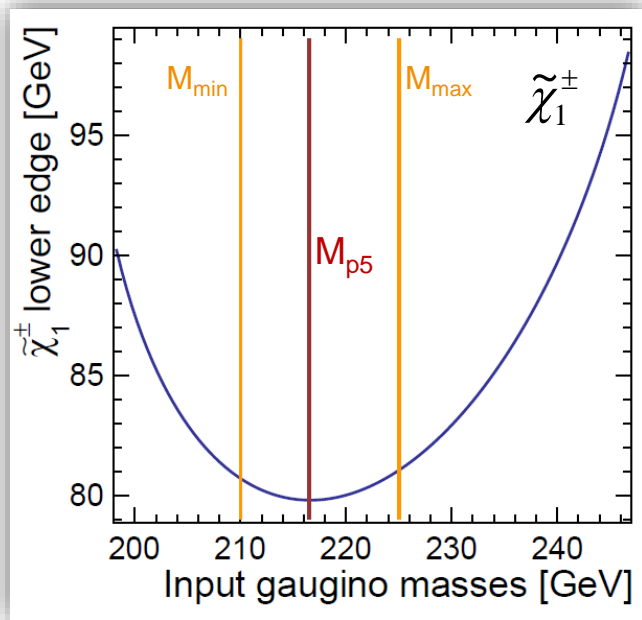
- > Canny's criteria for an optimal filter:
 - **J. F. Canny. A computational approach to edge detection.**
IEEE Trans. Pattern Analysis and Machine Intelligence, pages 679-698, 1986
 - **Good detection:** probability of obtaining a peak in the response must be high
 - **Localisation:** standard deviation 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 (σ) = **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:

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

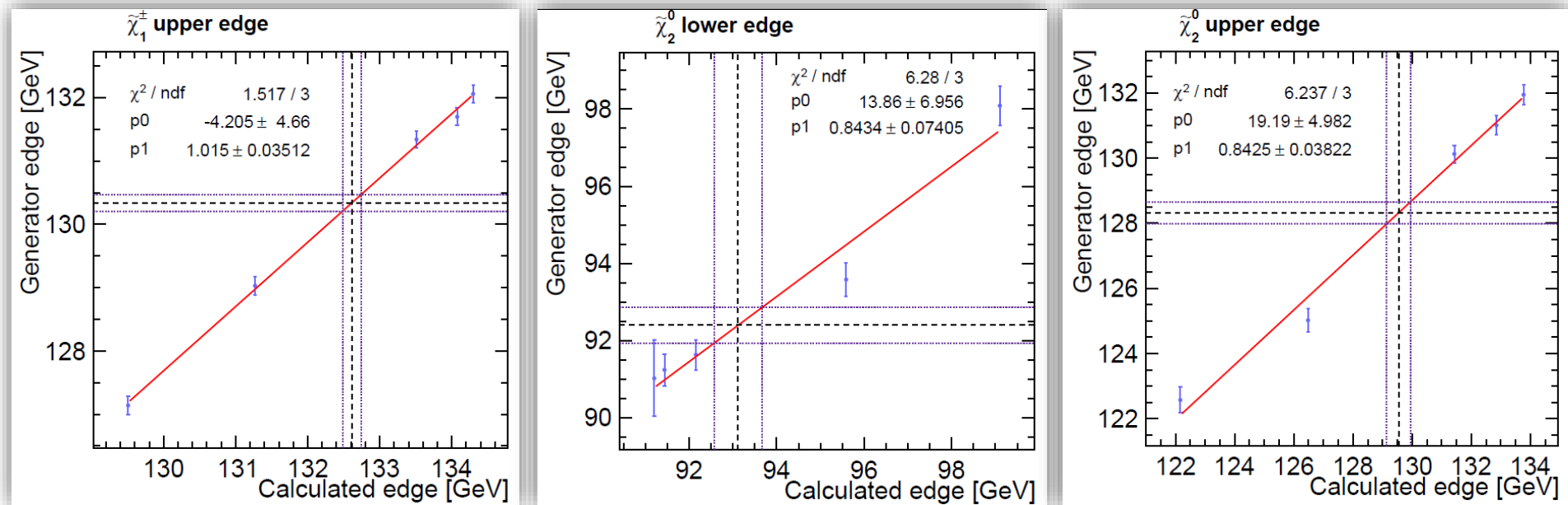


Edge Calibration II

➤ 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 ± 0.13	—	132.76	—	132.61 ± 0.13
$\tilde{\chi}_2^0$	92.395 ± 0.46	128.32 ± 0.34	93.09	129.92	93.11 ± 0.55	129.54 ± 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 ± 0.299	—	130.34 ± 0.13	—	130.39 ± 0.28
$\tilde{\chi}_2^0$	92.62 ± 0.24	127.82 ± 0.35	92.395 ± 0.46	128.32 ± 0.34	92.21 ± 0.29	128.64 ± 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 ± 0.78	—	132.77	—	132.79 ± 0.44
$\tilde{\chi}_2^0$	92.11 ± 0.31	128.99 ± 0.75	93.09	129.92	92.52 ± 1.23	130.67 ± 0.77

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

