



Fittino: Reverse engineering of supersymmetry

How to get the SUSY blueprint

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Inverse modelling

Often: What you want \neq what you get

Also often: Problems “easy” to solve in one direction, very difficult in opposite direction (one-way functions)

For many problems in science and engineering, getting what you want requires to follow difficult direction

Inverse modelling: observations \rightarrow model parameters

Examples:

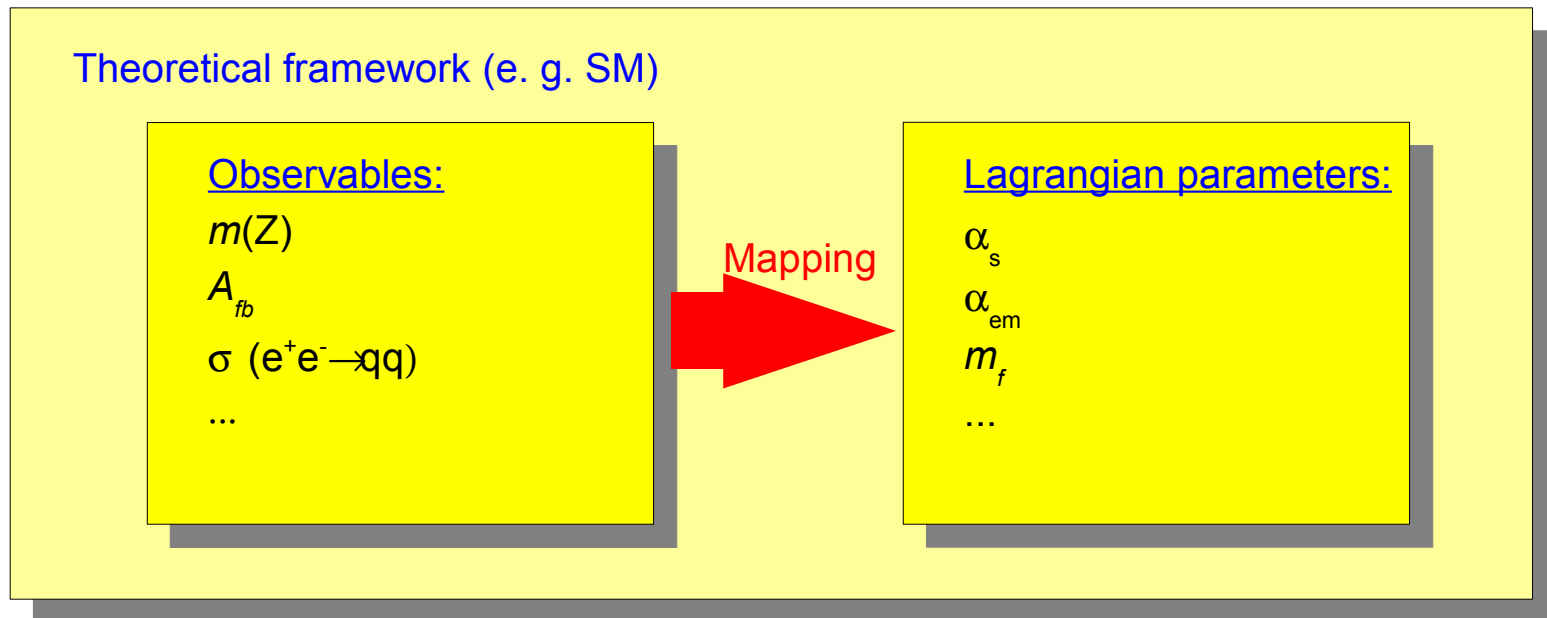
- Reverse engineering Technical device/software \rightarrow blueprint
- Scattering experiments Angles, energies, particle types \rightarrow particle/interaction properties
- Remote sensing E. g. multiple 1D/2D meas. \rightarrow 3D parameter map

Inverse problems are often “ill-posed” (Hadamard)

Inverse modelling in particle physics

Experimentalists provide: σ , BR, asymmetries, ...

Theorists provide: mapping: model parameters \rightarrow observables for various theories



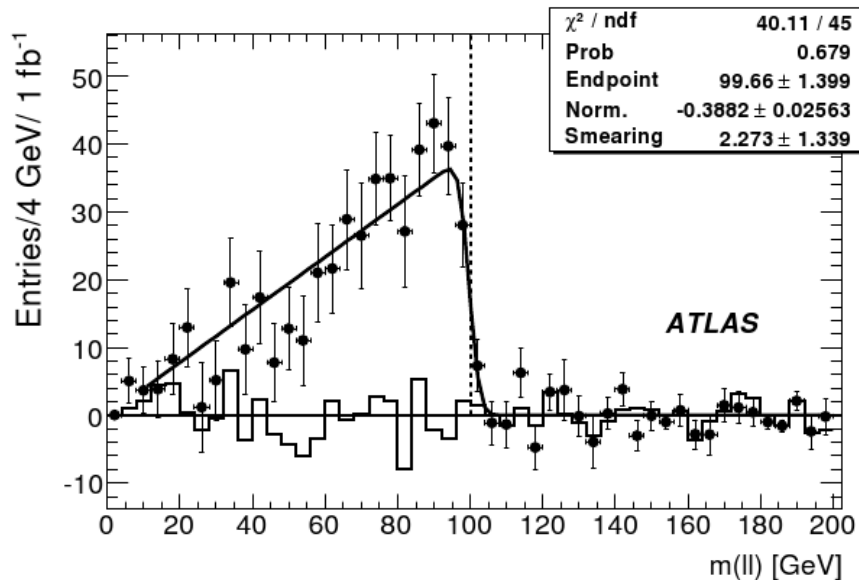
Need procedure to connect measurements to theory parameters for given theoretical framework

Supersymmetry (SUSY)

- Attractive candidate for extended SM of particle physics
- Remedies various shortcomings of SM
 - Hierarchy problem
 - No high-scale unification of gauge couplings
 - Lack of dark matter candidate
- SUSY as solution to SM problems only satisfactory if SUSY shows up **at the TeV scale**
 - Exciting prospects for LHC, ILC
 - Should get prepared for inverse modelling of supersymmetry

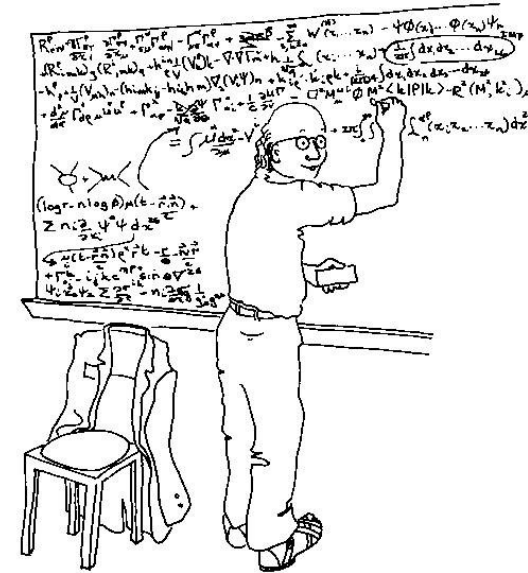
Inverse modelling of supersymmetry

Experiment:



Observables: m_{\parallel}^{\max} , m_h , ...

Theory:



Parameters: $\tan \beta$, μ , M_1 , ...

Questions to answer from measurements:

- What is the underlying (SUSY) model?
→ basically trial and error
- What are the values of its parameters?
→ needs sophisticated techniques

Inverse modelling of supersymmetry

At tree level, some sectors (e. g. chargino, chargino+neutralino) can be treated separately.

At loop level, in principle every observable depends on every parameter.

Complicated mutual dependence of the various parameters.

Approximate picture (not quite correct since non-linear mapping):

$$\begin{bmatrix} P_1 \\ P_2 \\ \vdots \end{bmatrix} = \begin{bmatrix} \square & & & 0 \\ & \square & & \\ & & \square & \\ 0 & & & \ddots \end{bmatrix} \begin{bmatrix} O_1 \\ O_2 \\ \vdots \end{bmatrix}$$

Tree level

$$\begin{bmatrix} P_1 \\ P_2 \\ \vdots \end{bmatrix} = \begin{bmatrix} \square & & & \neq 0 \\ & \square & & \\ & & \square & \\ \neq 0 & & & \ddots \end{bmatrix} \begin{bmatrix} O_1 \\ O_2 \\ \vdots \end{bmatrix}$$

Loop level

Iterative approach

Experiment:

- Measured observables O_i^m
- Errors ΔO_i^m

Output:

- SUSY parameters P_i
- Full error matrix V_{ij}

Tree-level formulae:

Rough estimates for:

- Parameters P_i
- Errors ΔP_i

SUSY calculation package:

- Calculated observables O_i^c
(including loop corrections)
- Full error matrix V_{ij}

χ^2 fit:
vary P

Compare

Programs

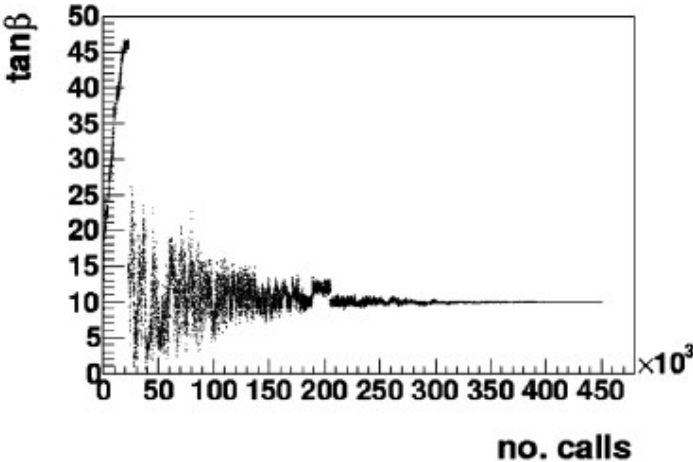
Several programs available which allow reconstruction of SUSY parameters from measurements:

- **Sfitter** (R. Lafaye, T. Plehn, M. Rauch, D. Zerwas)
- **Fittino** (P. Bechtle, K. Desch, M. Uhlenbrock, P. W.)
<http://www-flc.desy.de/fittino>
- **Gfitter** (H. Flücher, M. Goebel, J. Haller, A. Höcker, K. Mönig, J. Stelzer)
<http://gfitter.desy.de> (currently only SM and 2HDM)

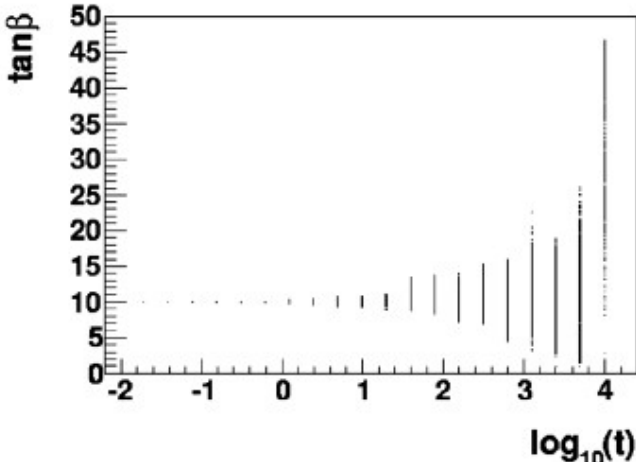
Fittino

- C++ package reconstructing SUSY parameters through χ^2 minimisation (using full correlation information)
- Currently supported SUSY models:
mSUGRA, GMSB, AMSB, MSSM24, NMSSM
- χ^2 minimisation using MINUIT or simulated annealing
- Calculation of likelihood maps using Markov chain Monte Carlo technique
- Theory predictions from SPheno (W. Porod) and Mastercode (Buchmüller *et al.*)

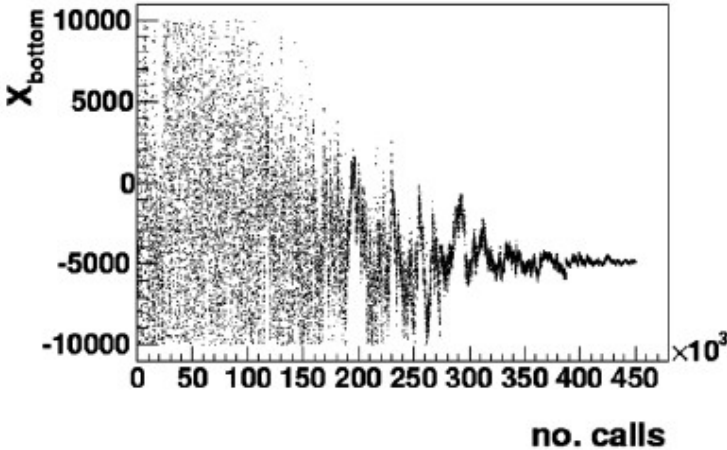
Simulated annealing in action



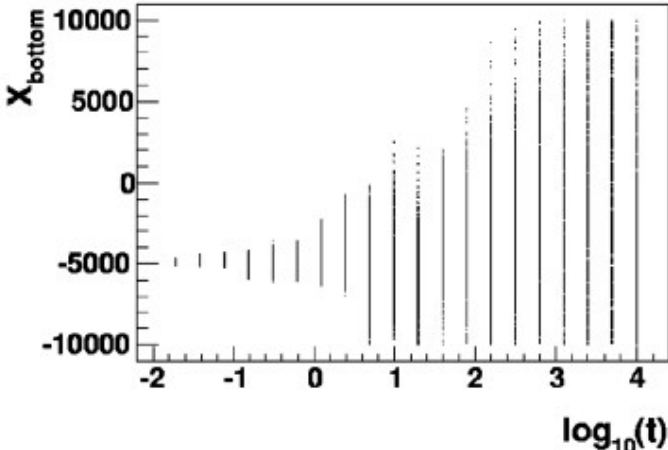
(a)



(b)



(c)



(d)

Markov chain Monte Carlo

Markov chain = sequence of points x_i ($i=1, \dots, n$) in parameter space
with associated likelihood

New point x_{n+1} randomly chosen according to proposal PDF is added to
chain if $\mathcal{L}(x_{n+1}) > \mathcal{L}(x_n)$

Otherwise it is accepted with probability $\mathcal{L}(x_{n+1})/\mathcal{L}(x_n)$

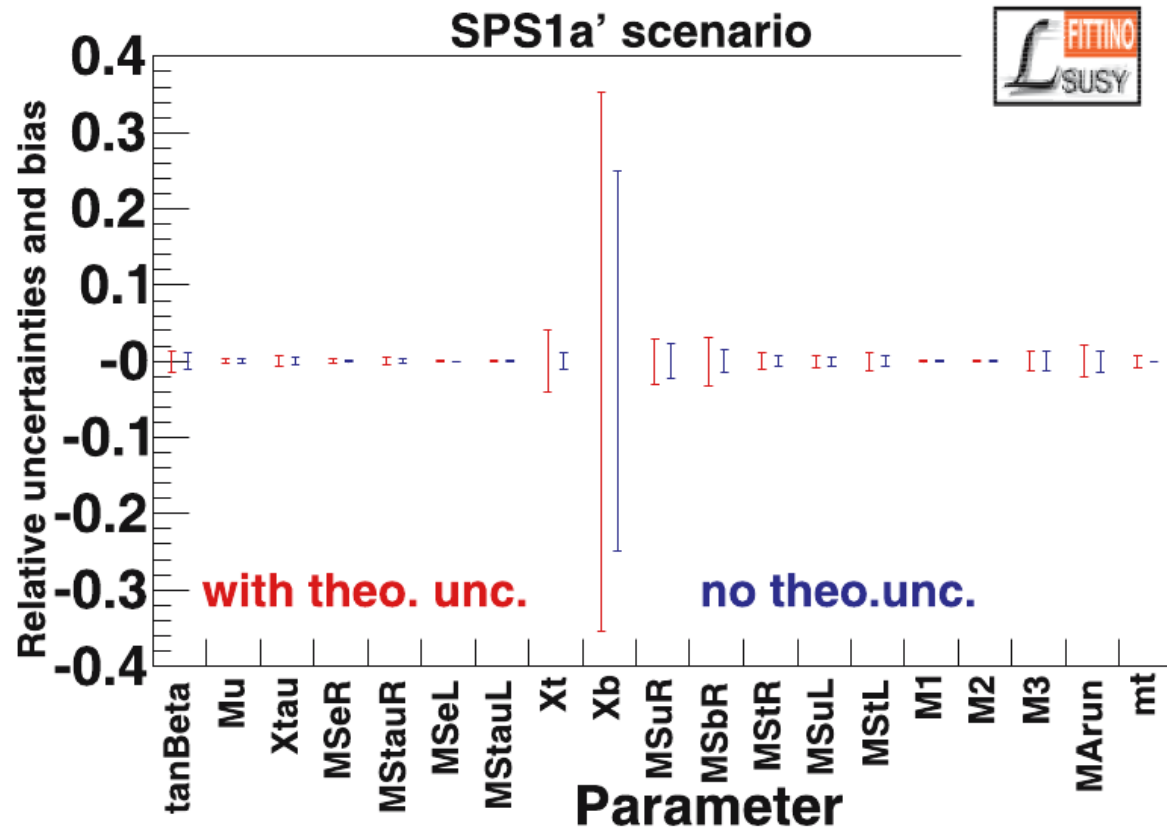
If proposal PDF is chosen properly, sampling density of points x_i in
Markov chain proportional to likelihood

History

Fittino originally developed to estimate the potential of combined LHC and ILC measurements (Eur. Phys. J. **C46**, 533-544, 2006)

Relatively far
future option

Recently we put
focus on presence
and near future



Available LE measurements

Wealth of “low“ energy (LE) measurements from past and present experiments:

- LEP, SLC
- B factories
- WMAP
- ...

They put **constraints on SUSY**

observable	meas. value	constraint	theo. uncert.
α_{em}	127.925	± 0.016	
α_S	0.1176	± 0.0020	
G_F (GeV ⁻²)	1.16637×10^{-5}	$\pm 0.00001 \times 10^{-5}$	
m_Z (GeV)	91.1875	± 0.0021	
m_W (GeV)	80.399	± 0.025	± 0.010
m_c (GeV)	1.27	± 0.11	
m_b (GeV)	4.20	± 0.17	
m_t (GeV)	172.4	± 1.2	
m_τ (GeV)	1.77684	± 0.00017	
m_h (GeV)	> 114.4		± 3.0
Γ_Z (MeV)	2495.2	± 2.3	± 1.0
Δa_μ	30.2×10^{-10}	$\pm 8.8 \times 10^{-10}$	$\pm 2.0 \times 10^{-10}$
σ_{had}^0 (nb)	41.540	± 0.037	
R_ℓ	20.767	± 0.025	
R_b	0.21629	± 0.00066	
R_c	0.1721	± 0.003	
A_{FB}^ℓ	0.01714	± 0.00095	
A_{FB}^b	0.0992	± 0.0016	
A_{FB}^c	0.0707	± 0.0035	
$A_\ell(\text{SLD})$	0.1513	± 0.0021	
$A_\ell(P_\tau)$	0.1465	± 0.0032	
A_b	0.923	± 0.020	
A_c	0.670	± 0.027	
$\sin^2 \theta_W^{\ell}(Q_{fb})$	0.2324	± 0.0012	
Ωh^2	0.1099	± 0.0062	± 0.012
$\text{BR}(B_d \rightarrow \mu^+ \mu^-)$		$< 2.3 \times 10^{-8}$	$\pm 0.01 \times 10^{-9}$
$\text{BR}(B_s \rightarrow \mu^+ \mu^-)$		$< 4.7 \times 10^{-8}$	$\pm 0.02 \times 10^{-8}$
$R(b \rightarrow s\gamma)$	1.117	$\pm 0.076 \pm 0.082$	± 0.050
$R(B \rightarrow \tau\nu)$	1.15	± 0.40	
$R(B \rightarrow X_s \ell\ell)$	0.99	± 0.32	
$R(K \rightarrow \mu\nu)$	1.008	± 0.014	
$R(K \rightarrow \pi\nu\bar{\nu})$		< 4.5	
$R(\Delta m_{B_s})$	1.11	± 0.01	± 0.32
$R(\Delta m_{B_s})/R(\Delta m_{B_d})$	1.09	± 0.01	± 0.16
$R(\Delta\epsilon_K)$	0.92	± 0.14	

SM+mSUGRA fit to LE measurements

Fit of α_{em} , α_s , G_F , m_Z , m_b , m_t , m_τ and mSUGRA parameters to LE measurements:

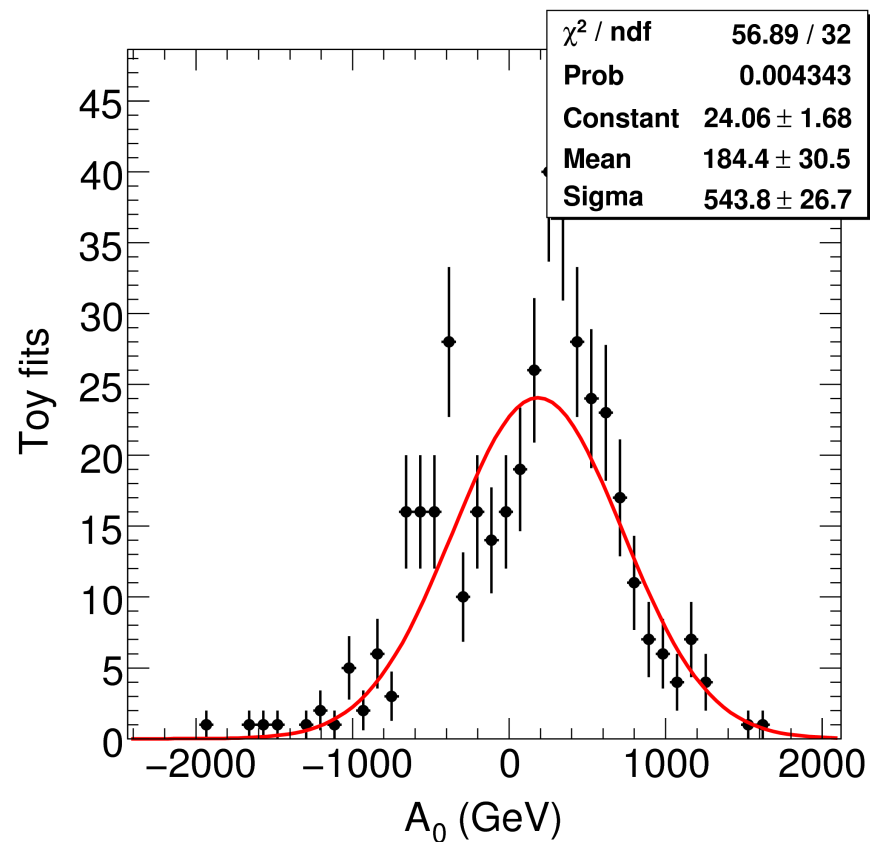
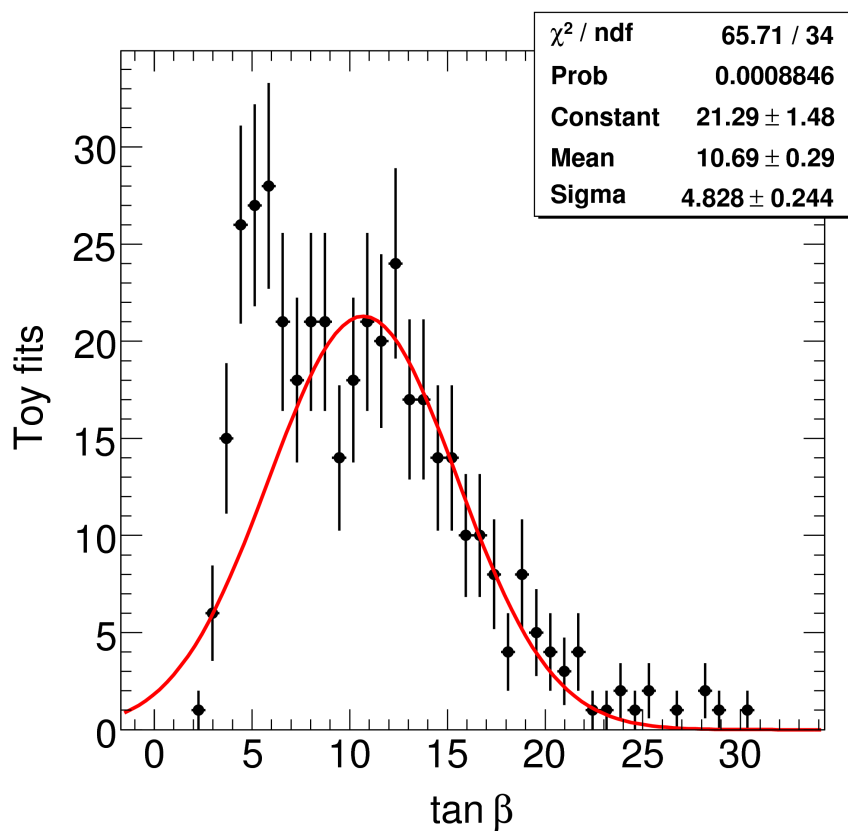
Best agreement for:

Parameter	Value	Uncertainty
$\tan \beta$	12.4	4.8
A_0 (GeV)	337	544
M_0 (GeV)	71	18
$M_{1/2}$ (GeV)	323	62

Uncertainties obtained from toy fits to observables smeared around nominal values for best fit parameters

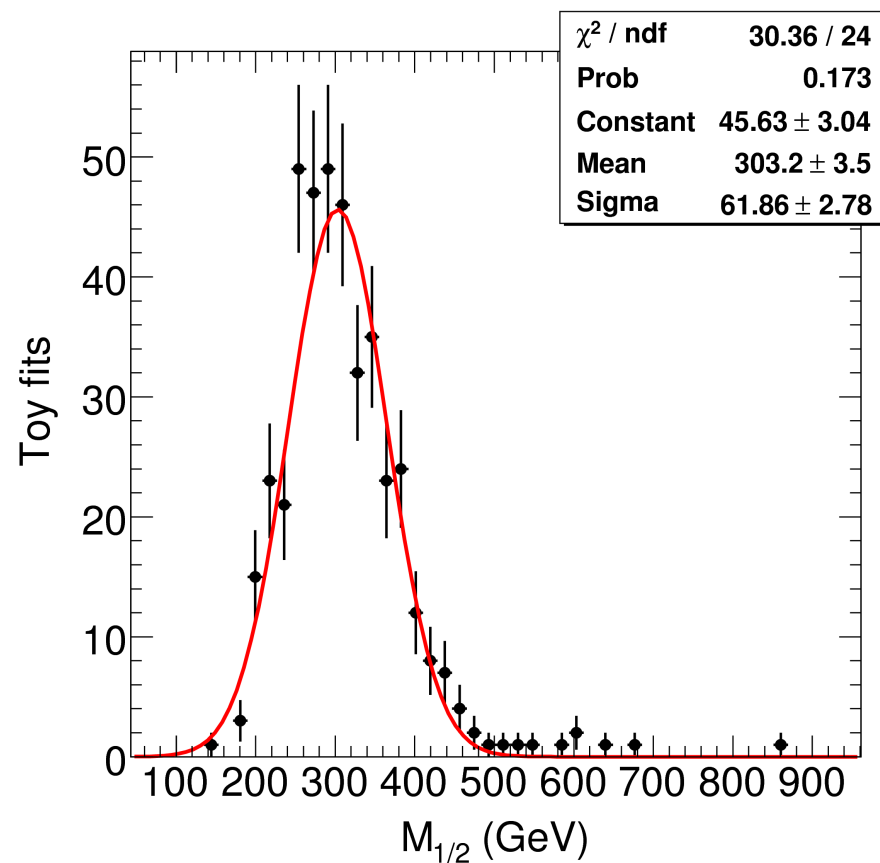
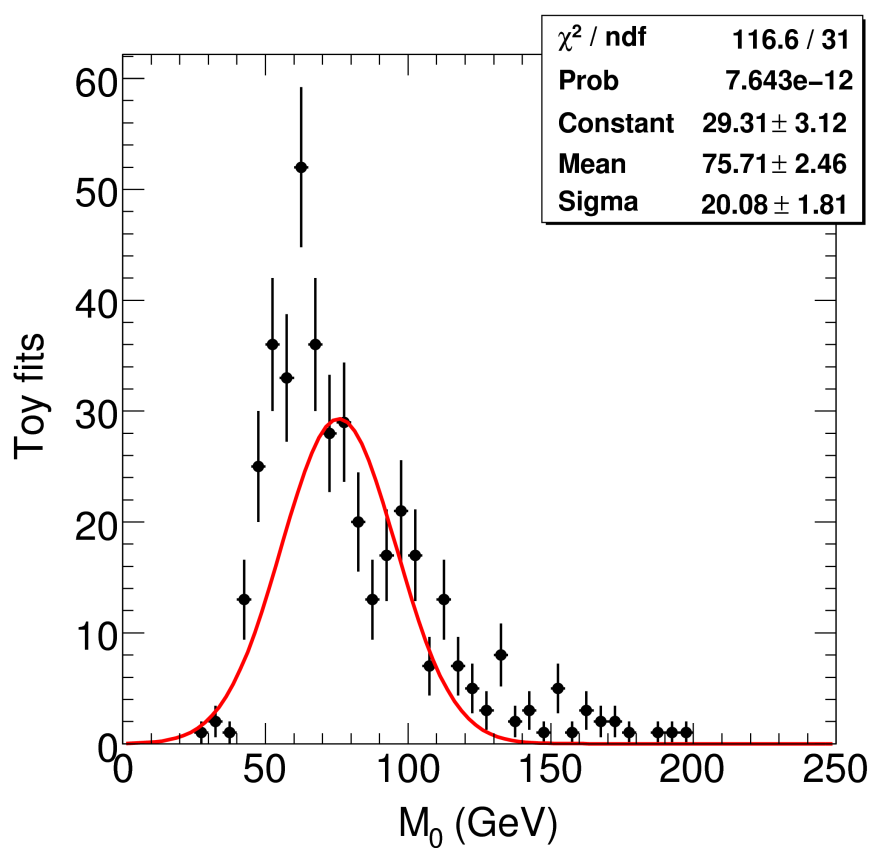
SM+mSUGRA fit to LE measurements

Parameter distributions for toy fits:



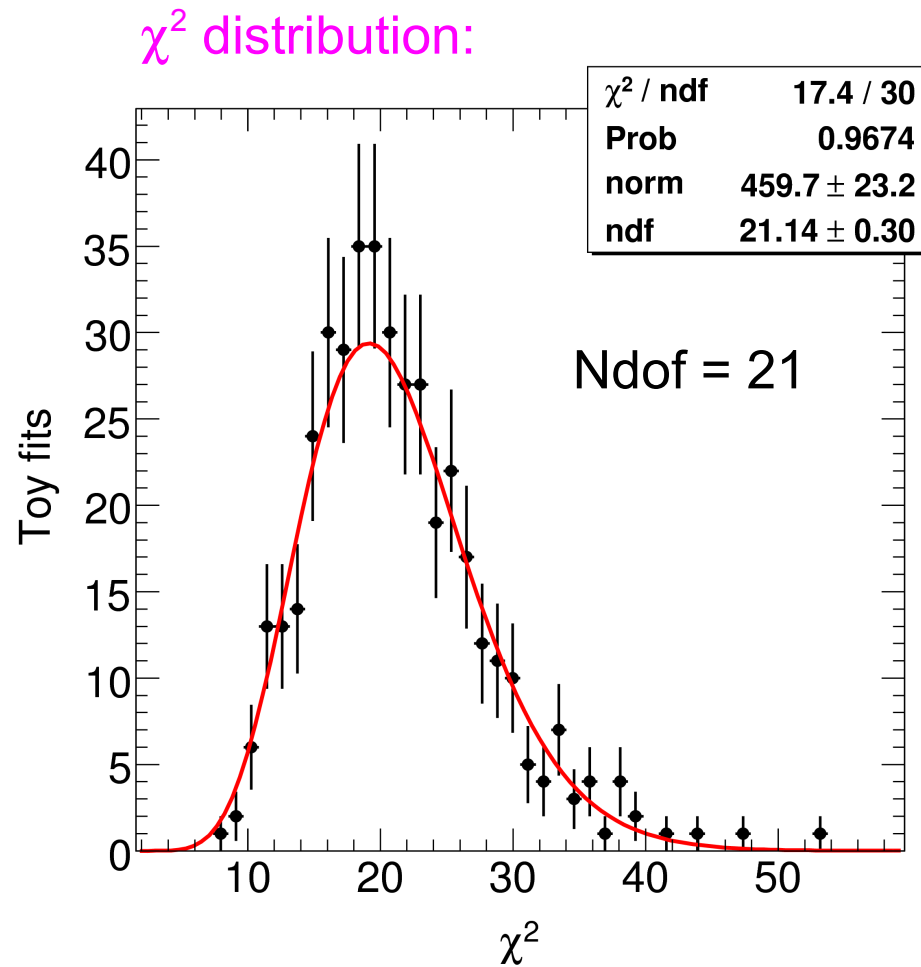
SM+mSUGRA fit to LE measurements

Parameter distributions for toy fits:

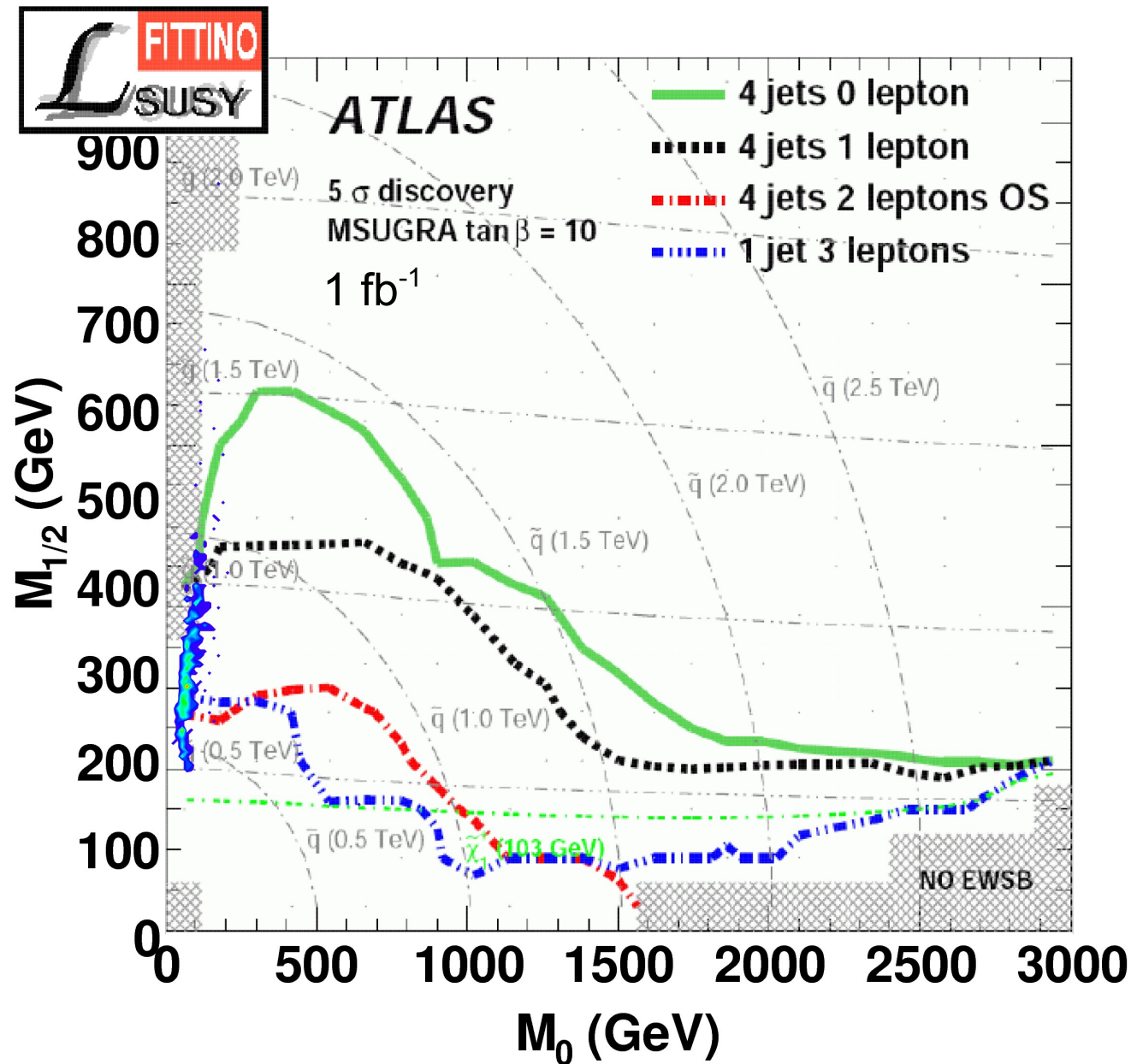


SM+mSUGRA fit to LE measurements

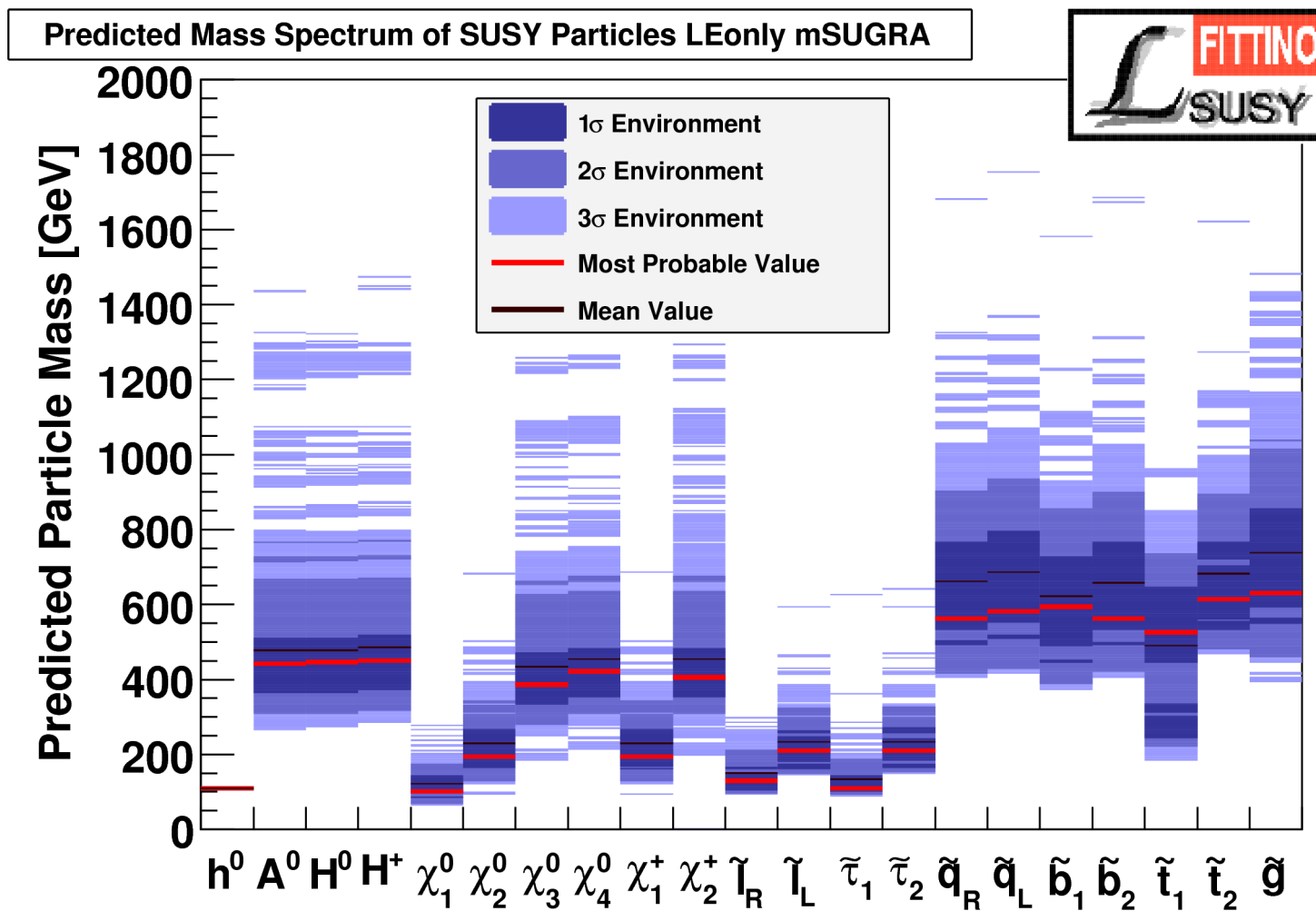
Distributions for toy fits:



Comparison: ATLAS discovery potential ↔ mSUGRA fit results

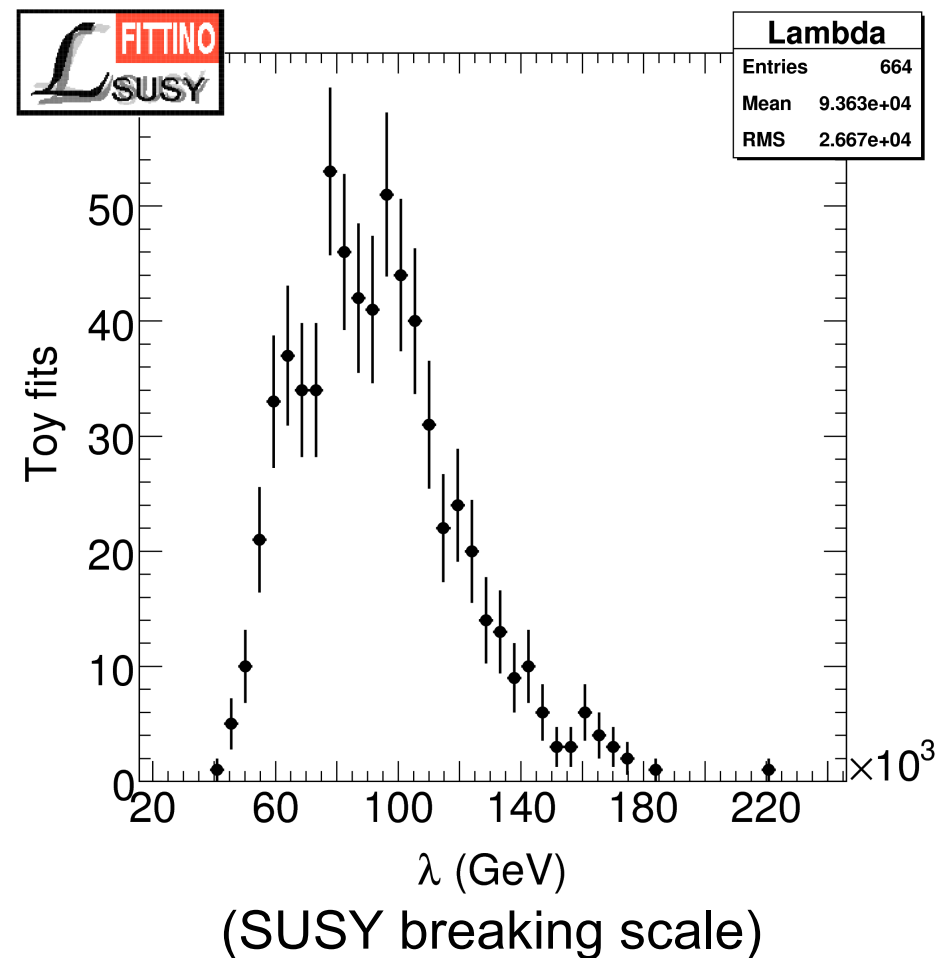
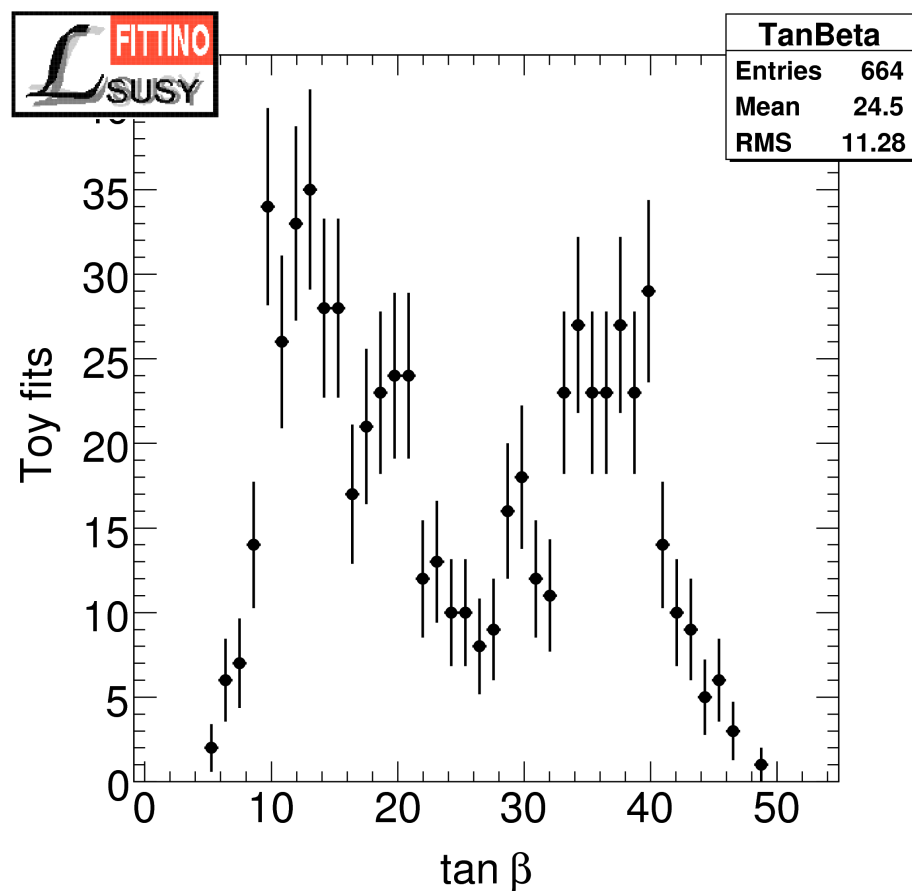


Predicted mass spectrum from mSUGRA fit



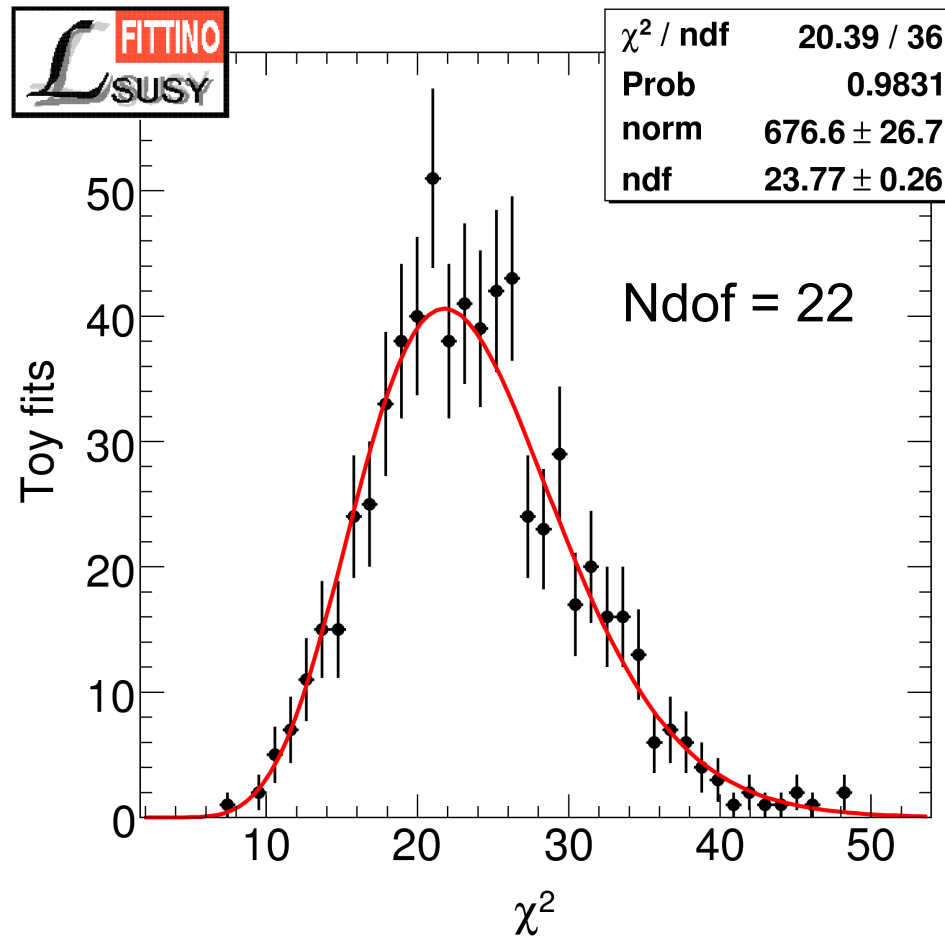
SM+GMSB fit to LE measurements

Parameter distributions for toy fits with observables smeared around nominal values for parameters of best fit:

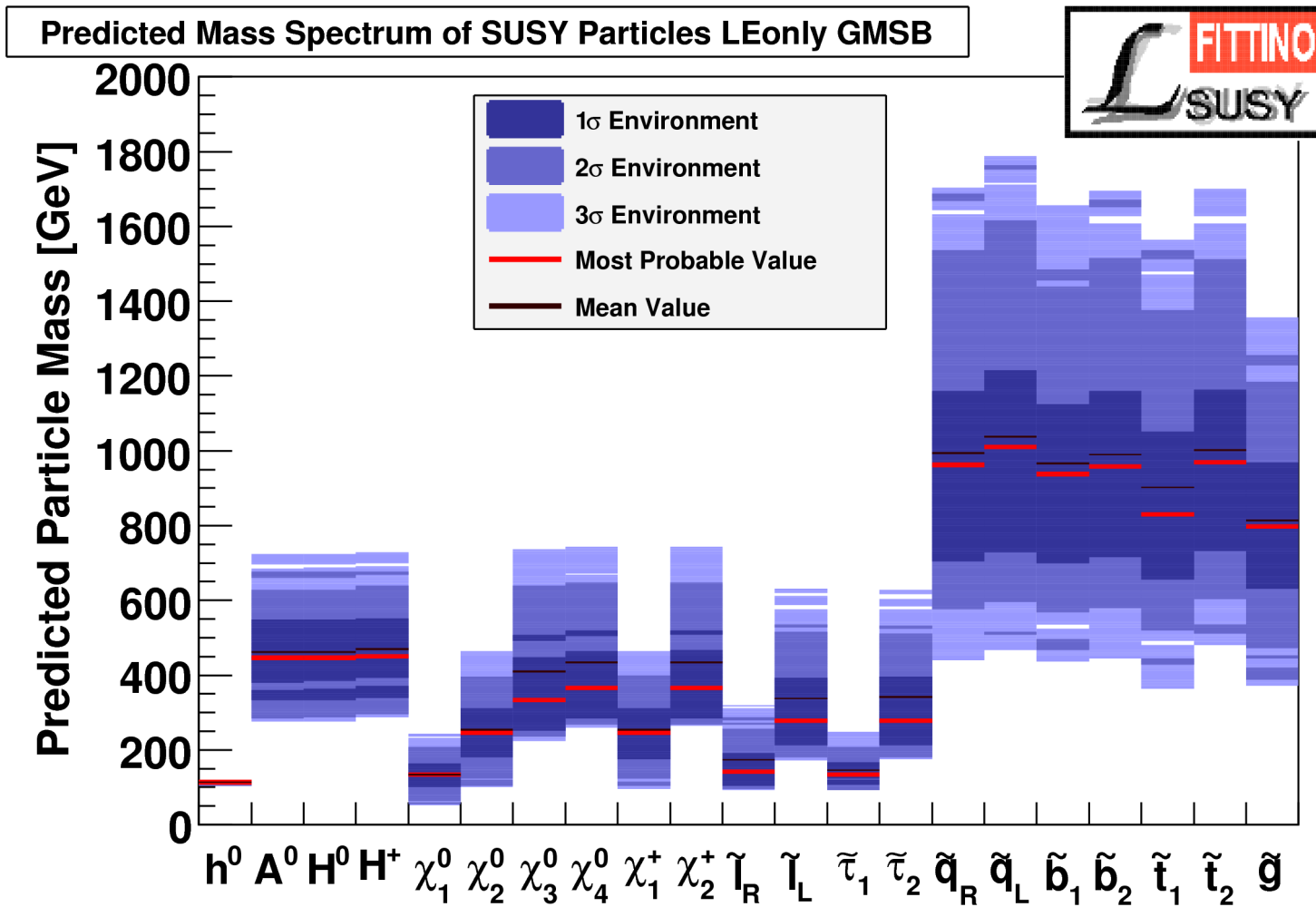


SM+GMSB fit to LE measurements

χ^2 distribution for toy fits:

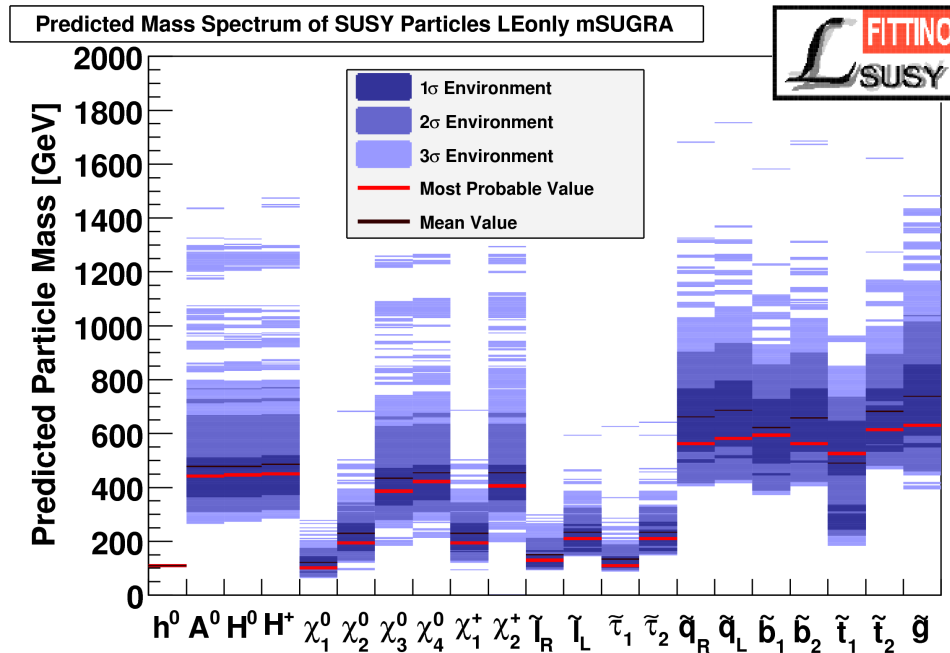


Predicted mass spectrum from SM+GMSB fit

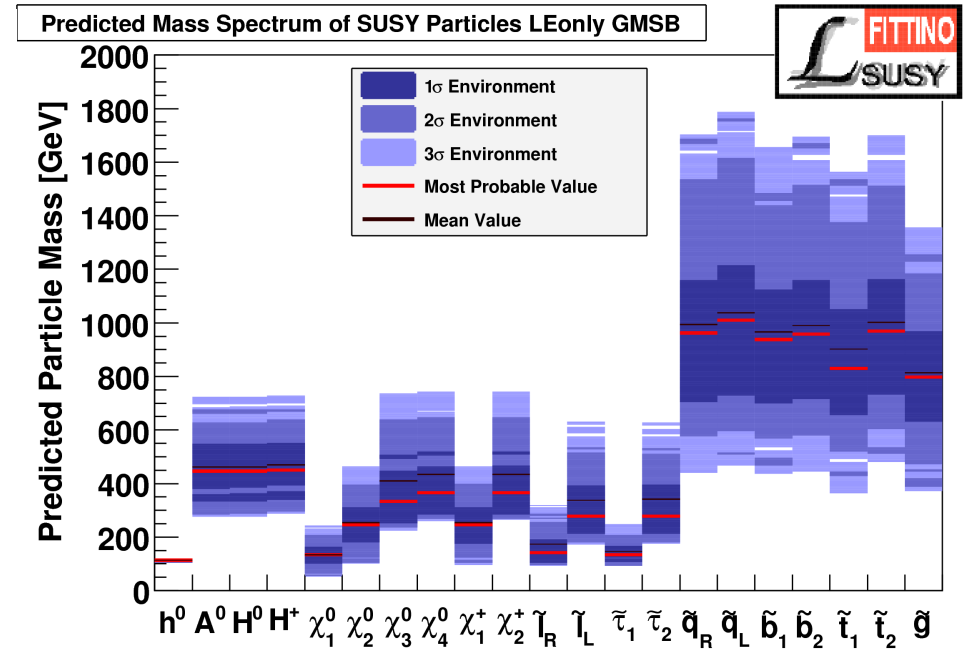


Comparison: mSUGRA ↔ GMSB spectrum

mSUGRA



GMSB



LHC “measurements”

LHC is just around the corner and will hopefully shower us with exciting new measurements.

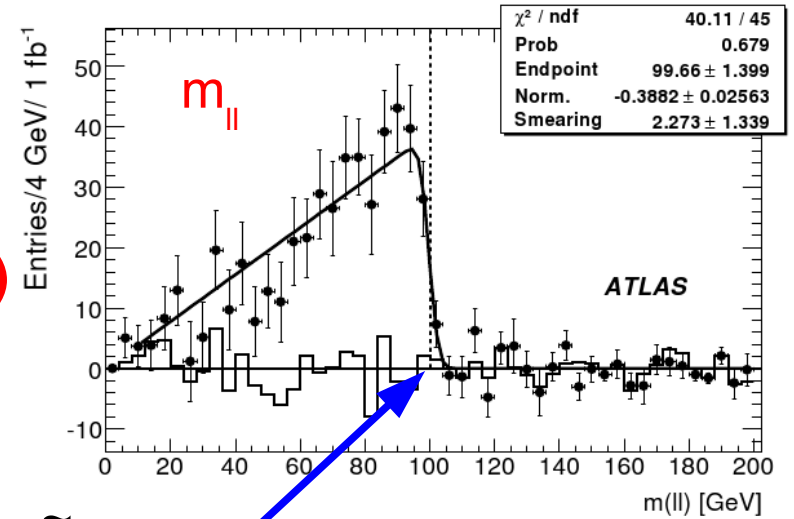
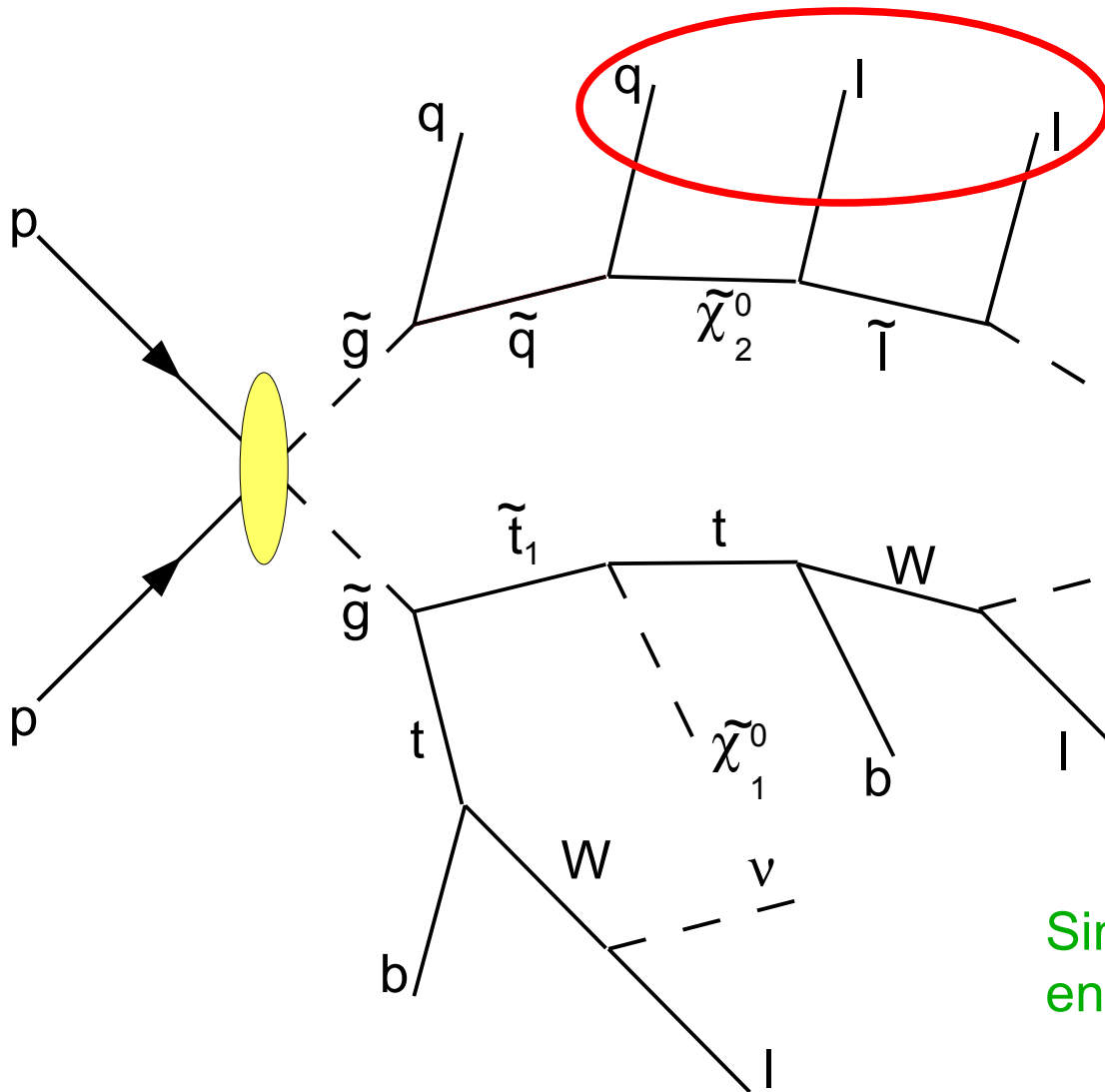
Case study for SPS1a assuming 1 fb⁻¹, 10 fb⁻¹ and 300 fb⁻¹:

Observable	Nominal Value	Uncertainty							
		1 fb ⁻¹	10 fb ⁻¹	300 fb ⁻¹	LES ₁	LES _{10,300}	JES ₁	JES _{10,300}	syst.
m_h	109.1		1.4	0.1		0.1			
m_t	170.9	1.1*	0.05	0.01			1.5*	1.0	
$m_{\tilde{\chi}_1^\pm}$	179.9			11.4				1.8	
$m_{\tilde{\ell}_L} - m_{\tilde{\chi}_1^0}$	105.4			1.7		0.1			6.0
$m_{\tilde{g}} - m_{\tilde{\chi}_1^0}$	510.2		13.7	2.5				5.1	10.0
$m_{\tilde{q}_R} - m_{\tilde{\chi}_1^0}$	454.0	19.6	6.2	1.1			22.7	4.5	10.0
$\langle m_{\tilde{g}} - m_{\tilde{b}_{1,2}} \rangle$	522.6		5.4					5.2	
$m_{\tilde{g}} - m_{\tilde{b}_1}$	89.0			1.5				0.9	
$m_{\tilde{g}} - m_{\tilde{b}_2}$	56.7			2.5				0.6	
$m_{\ell\ell}^{\max} = \epsilon_1(m_{\tilde{\chi}_1^0}, m_{\tilde{\chi}_2^0}, m_{\tilde{\ell}_R})$	80.2	1.7	0.5	0.03	0.16	0.08			
$m_{\ell\ell}^{\max} = \epsilon_1(m_{\tilde{\chi}_1^0}, m_{\tilde{\chi}_4^0}, m_{\tilde{\ell}_L})$	279.1		12.6	2.3		0.28			
$m_{\tau\tau}^{\max} = \epsilon_1(m_{\tilde{\chi}_1^0}, m_{\tilde{\chi}_2^0}, m_{\tilde{\tau}_1})$	83.2	12.6	4.0	0.73			4.2	0.8	5.7
$m_{\ell\ell q}^{\max} = \epsilon_1(m_{\tilde{\chi}_1^0}, m_{\tilde{q}_L}, m_{\tilde{\chi}_2^0})$	454.3	13.9	4.2	1.4			22.7	4.5	
$m_{\ell q}^{\text{low}} = \epsilon_1(m_{\tilde{\ell}_R}, m_{\tilde{q}_L}, m_{\tilde{\chi}_2^0})$	324.2	7.6	3.5	0.9			16.2	3.2	
$m_{\ell q}^{\text{high}} = \epsilon_2(m_{\tilde{\chi}_1^0}, m_{\tilde{\chi}_2^0}, m_{\tilde{\ell}_R}, m_{\tilde{q}_L})$	398.3	5.2	4.5	1.0			19.9	4.0	
$m_{\ell\ell q}^{\text{thrs}} = \epsilon_3(m_{\tilde{\chi}_1^0}, m_{\tilde{\chi}_2^0}, m_{\tilde{\ell}_R}, m_{\tilde{q}_L})$	216.2	26.5	4.8	1.6			10.8	2.2	
$m_{\ell\ell b}^{\text{thrs}} = \epsilon_3(m_{\tilde{\chi}_1^0}, m_{\tilde{\chi}_2^0}, m_{\tilde{\ell}_R}, m_{\tilde{b}_1})$	196.4		19.7	3.6				2.0	
$m_{tb}^w = \epsilon_4(m_t, m_{\tilde{t}_1}, m_{\tilde{\chi}_j^\pm}, m_{\tilde{g}}, m_{\tilde{b}_i})$	360.9	43.0	13.6	2.5			18.0	3.6	
$\frac{\text{BR}(\tilde{\chi}_2^0 \rightarrow \tilde{\ell}\ell) \times \text{BR}(\tilde{\ell} \rightarrow \tilde{\chi}_1^0 \ell)}{\text{BR}(\tilde{\chi}_2^0 \rightarrow \tilde{\tau}_1 \tau) \times \text{BR}(\tilde{\tau}_1 \rightarrow \tilde{\chi}_1^0 \tau)}$	0.08	0.009	0.003	0.001					0.008
$\frac{\text{BR}(\tilde{g} \rightarrow \tilde{b}_2 b) \times \text{BR}(\tilde{b}_2 \rightarrow \tilde{\chi}_2^0 b)}{\text{BR}(\tilde{g} \rightarrow \tilde{b}_1 b) \times \text{BR}(\tilde{b}_1 \rightarrow \tilde{\chi}_2^0 b)}$	0.16			0.078					

Mass reconstruction at LHC (RPC SUSY)

No mass peaks due to 2 escaping LSPs

Most important **mass information source**:



Endpoint sensitive to SUSY masses:

$$(m_{ll}^2)^{\text{edge}} = \frac{(m_{\tilde{\chi}_2^0}^2 - m_{\tilde{l}_R}^2)(m_{\tilde{l}_R}^2 - m_{\tilde{\chi}_1^0}^2)}{m_{\tilde{l}_R}^2}$$

Similar formulae for other endpoints, e. g. m_{llq}^{min} , m_{llq}^{max} , ...

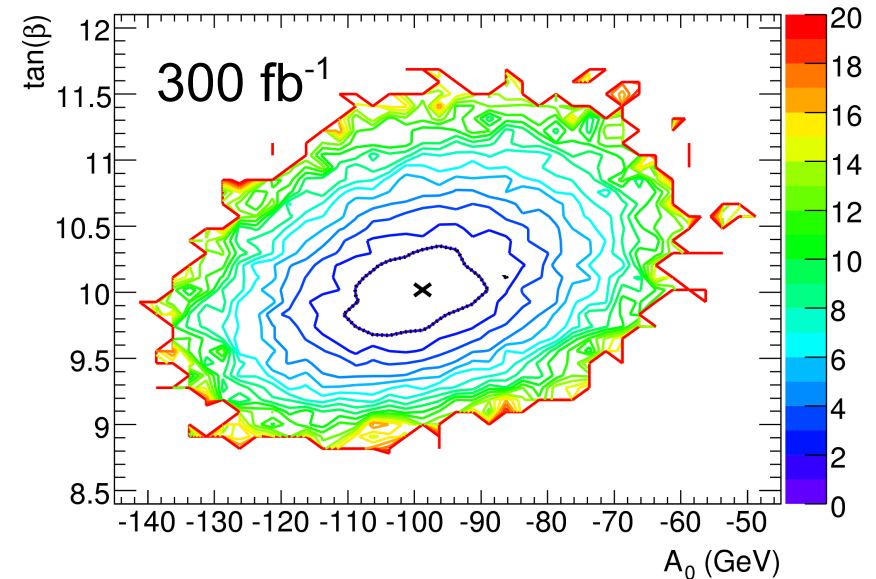
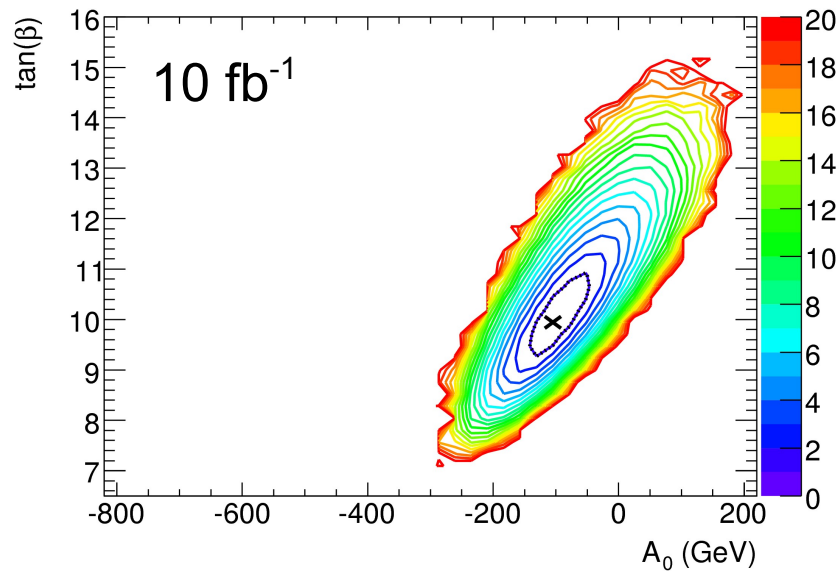
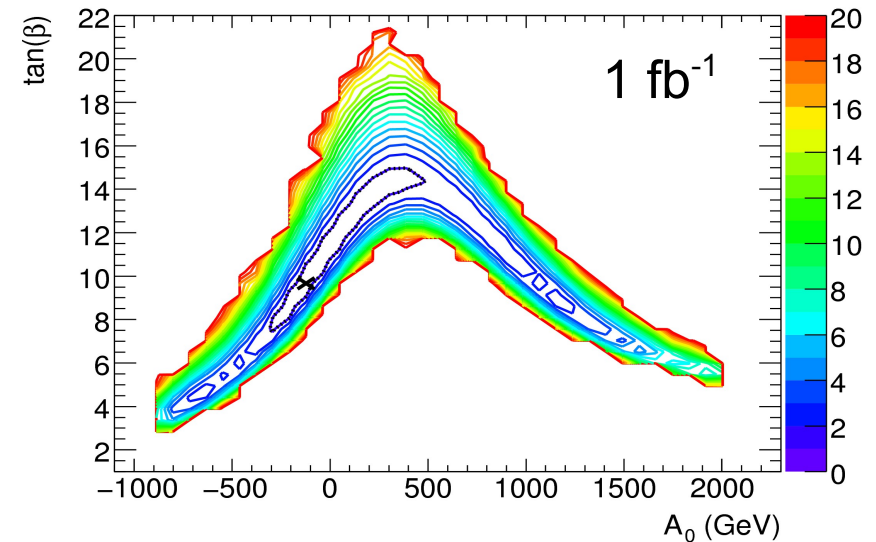
Likelihood maps from LHC “measurements”

Likelihood maps for mSUGRA parameters, $\text{sign}(\mu)$ fixed to +

Plots show contours of $2 \ln(\mathcal{L}_{\text{max}}/\mathcal{L})$

Note different scales!

SPS1a: $\tan \beta = 10$, $A_0 = -100$ GeV



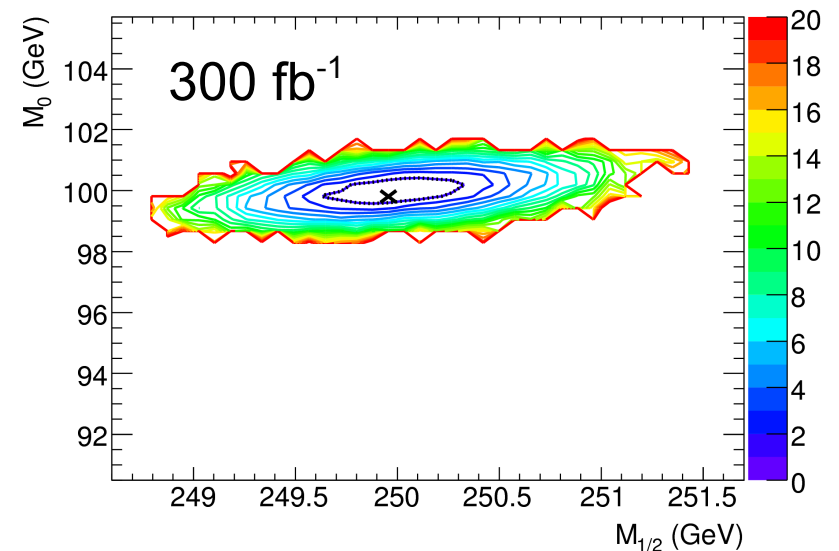
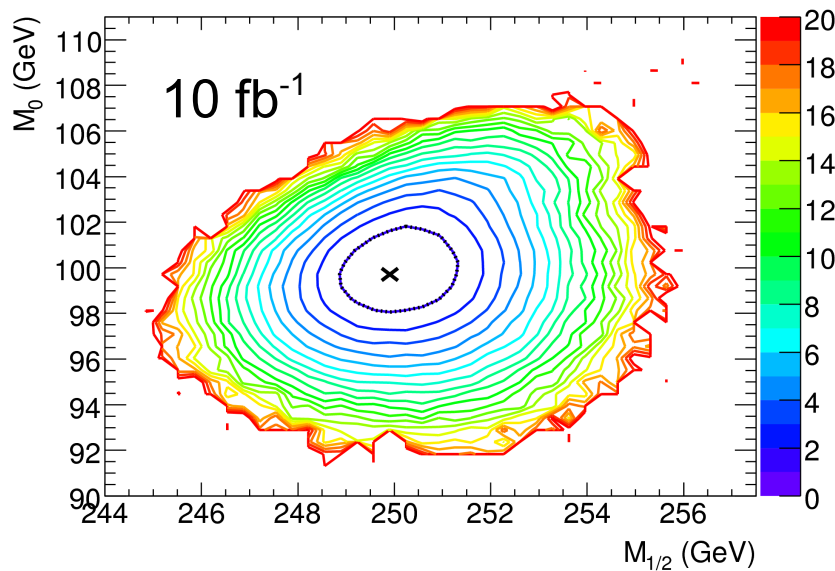
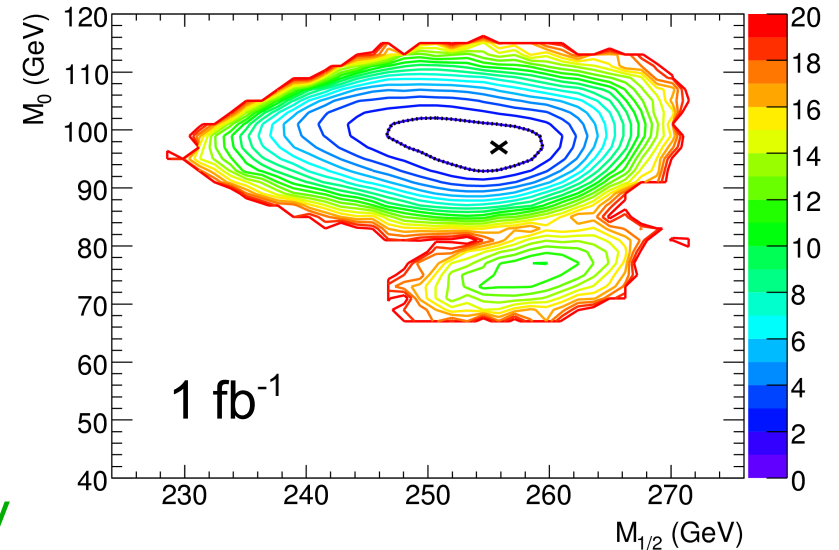
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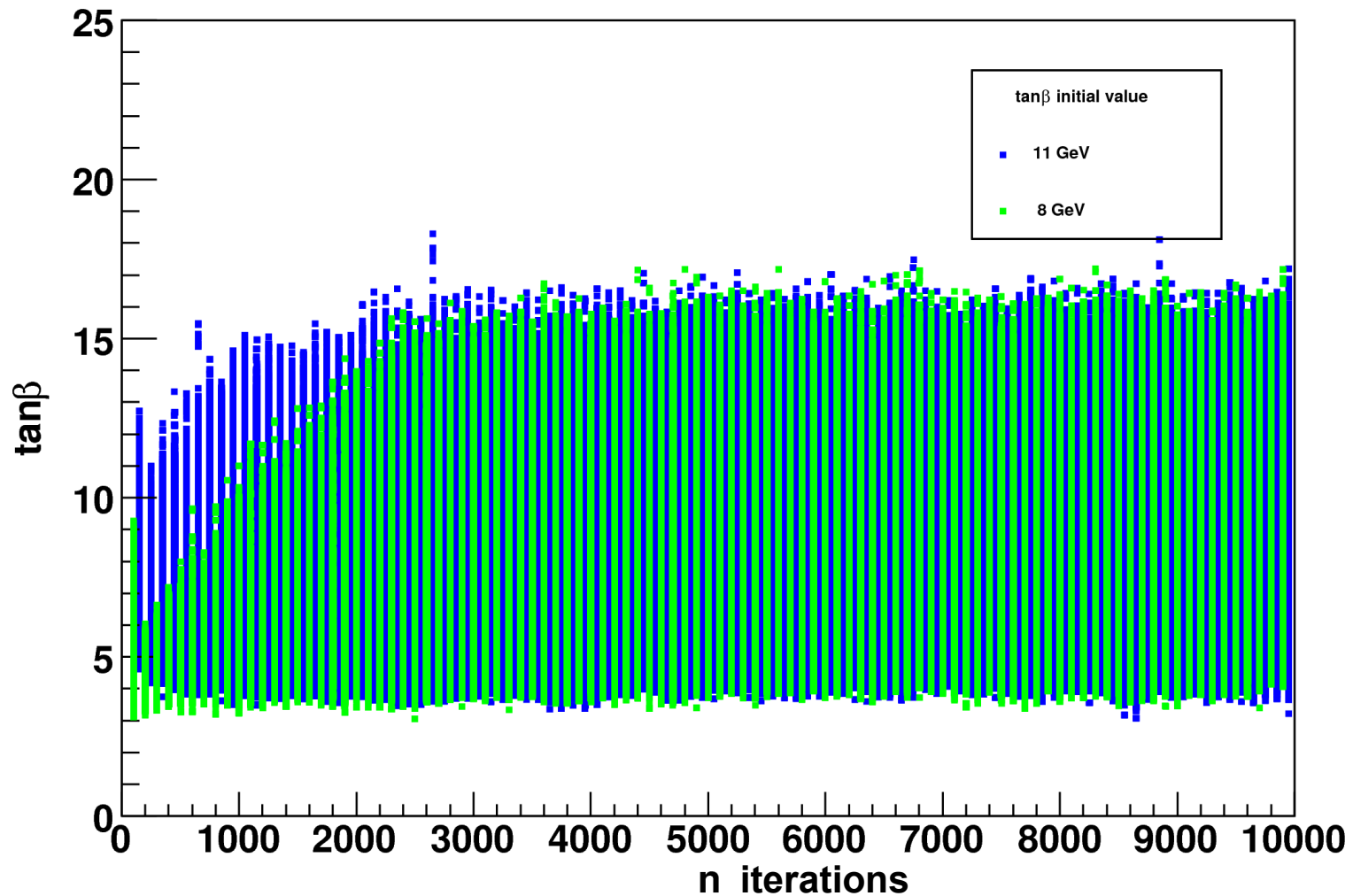
Note different scales!

SPS1a: $M_0 = 100 \text{ GeV}$, $M_{1/2} = 250 \text{ GeV}$



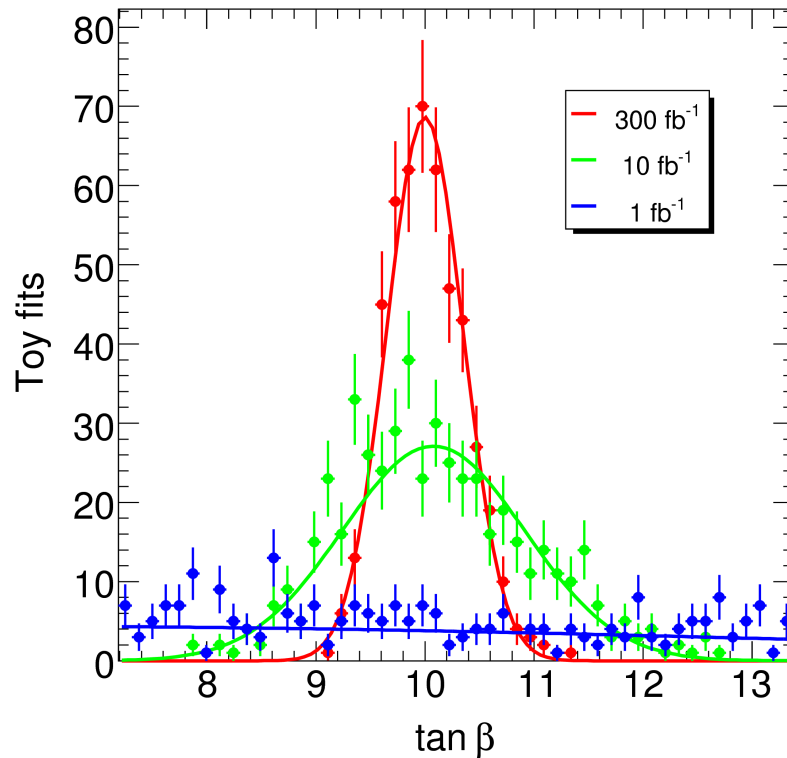
Caveat

Markov chains need sufficient number of iterations to settle down, i. e. results become independent of start values



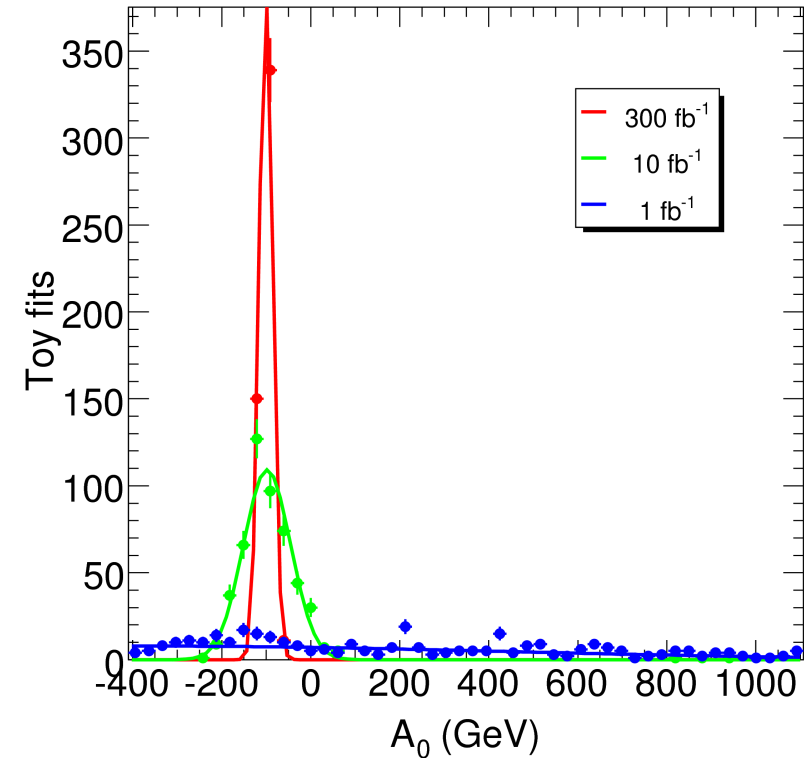
mSUGRA parameter uncertainties from LHC

mSUGRA parameter distributions for toy fits:



Luminosity	Uncertainty
1 fb-1	3.7 (37 %)
10 fb-1	0.8 (8 %)
300 fb-1	0.4 (4 %)

SPS1a: $\tan \beta = 10$

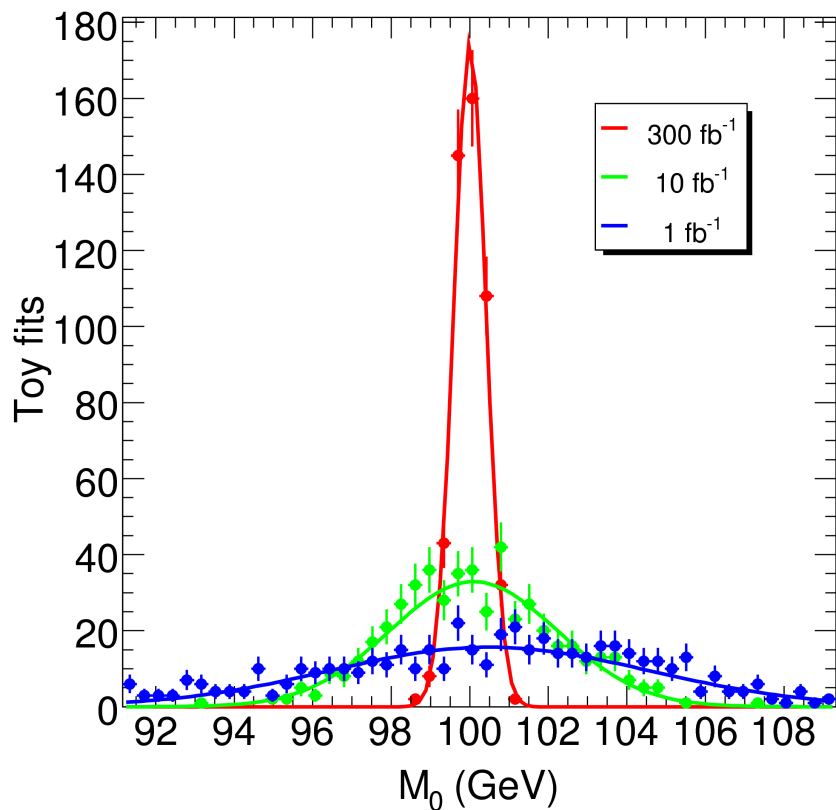


Luminosity	Uncertainty
1 fb-1	742 (742 %)
10 fb-1	53 (53 %)
300 fb-1	11 (11 %)

SPS1a: $A_0 = -100$ GeV

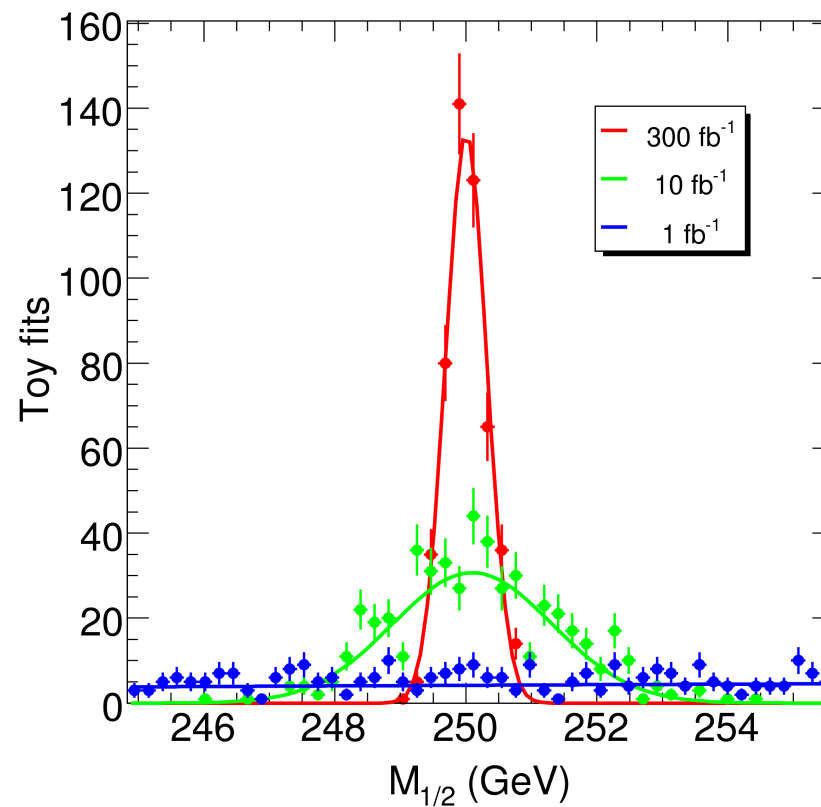
mSUGRA parameter uncertainties from LHC

mSUGRA parameter distributions for toy fits:



Luminosity	Uncertainty
1 fb ⁻¹	4.2 (4.2 %)
10 fb ⁻¹	2.1 (2.1 %)
300 fb ⁻¹	0.39 (0.4 %)

SPS1a: $M_0 = 100$ GeV



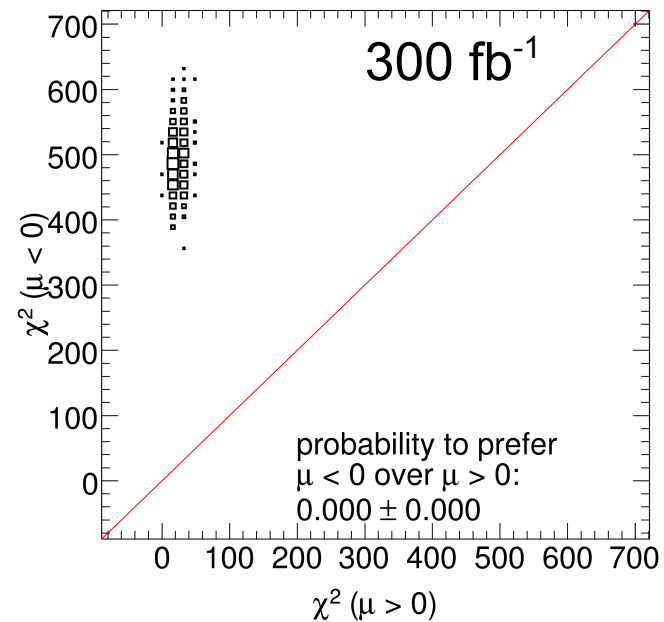
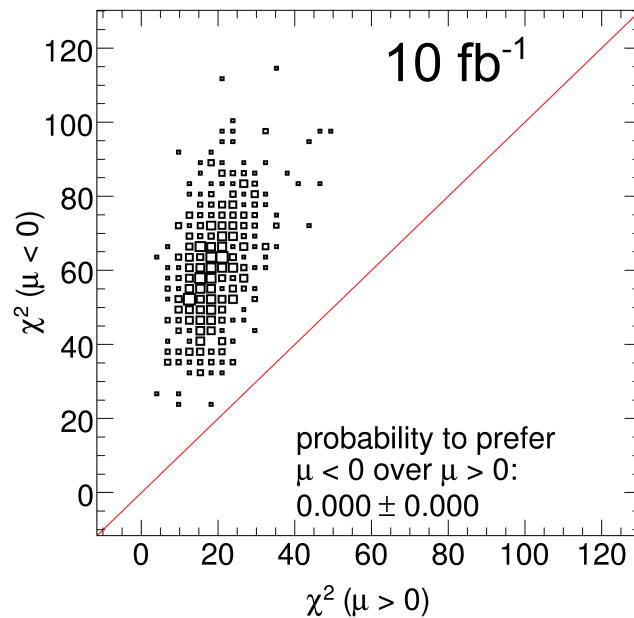
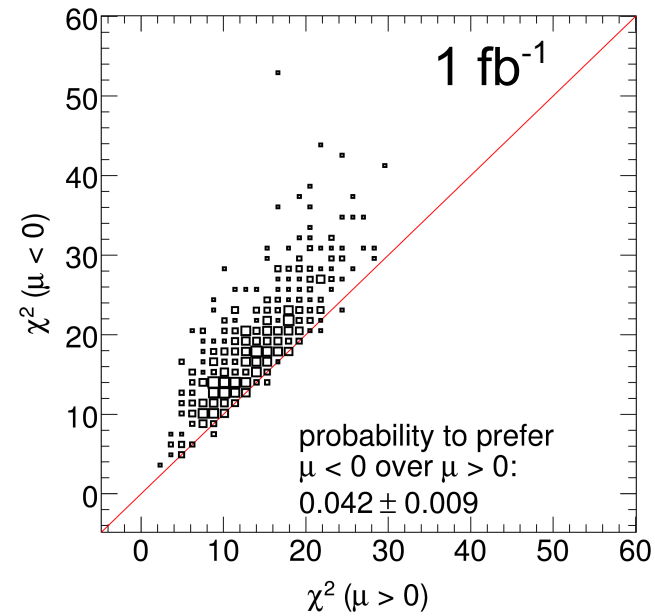
Luminosity	Uncertainty
1 fb ⁻¹	6.7 (2.7 %)
10 fb ⁻¹	1.2 (0.5 %)
300 fb ⁻¹	0.30 (0.1 %)

SPS1a: $M_{1/2} = 250$ GeV

$\mu > 0$ vs. $\mu < 0$ from LHC “measurements”

Performed two fits (with $\mu > 0$ and $\mu < 0$) for every toy data set smeared around best fit values and compared χ^2 values

SPS1a: $\mu > 0$



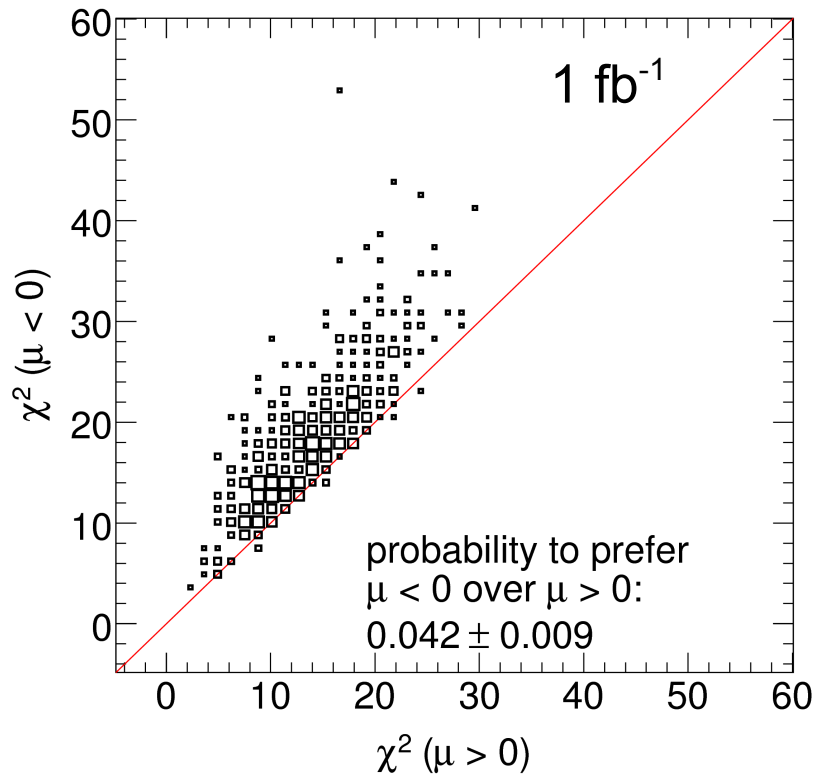
Combination LHC+LE

$\tan \beta$	Luminosity	Uncertainty LHC	Uncertainty LHC+LE
	1 fb-1	3.7 (41 %)	2.5 (25 %)
	10 fb-1	0.8 (8 %)	0.8 (8 %)
	300 fb-1	0.4 (4 %)	0.3 (3 %)
A_0 (GeV)	Luminosity	Uncertainty LHC	Uncertainty LHC+LE
	1 fb-1	742 (742 %)	169 (169 %)
	10 fb-1	53 (53 %)	48 (48 %)
	300 fb-1	11 (11 %)	12 (12 %)
M_0 (GeV)	Luminosity	Uncertainty LHC	Uncertainty LHC+LE
	1 fb-1	4.2 (4.2 %)	3.3 (3.3 %)
	10 fb-1	2.1 (2.1 %)	1.9 (1.9 %)
	300 fb-1	0.39 (0.4 %)	0.44 (0.4 %)
$M_{1/2}$ (GeV)	Luminosity	Uncertainty LHC	Uncertainty LHC+LE
	1 fb-1	6.7 (2.7 %)	4.9 (2.0 %)
	10 fb-1	1.2 (0.5 %)	1.1 (0.4 %)
	300 fb-1	0.30 (0.1 %)	0.32 (0.1 %)

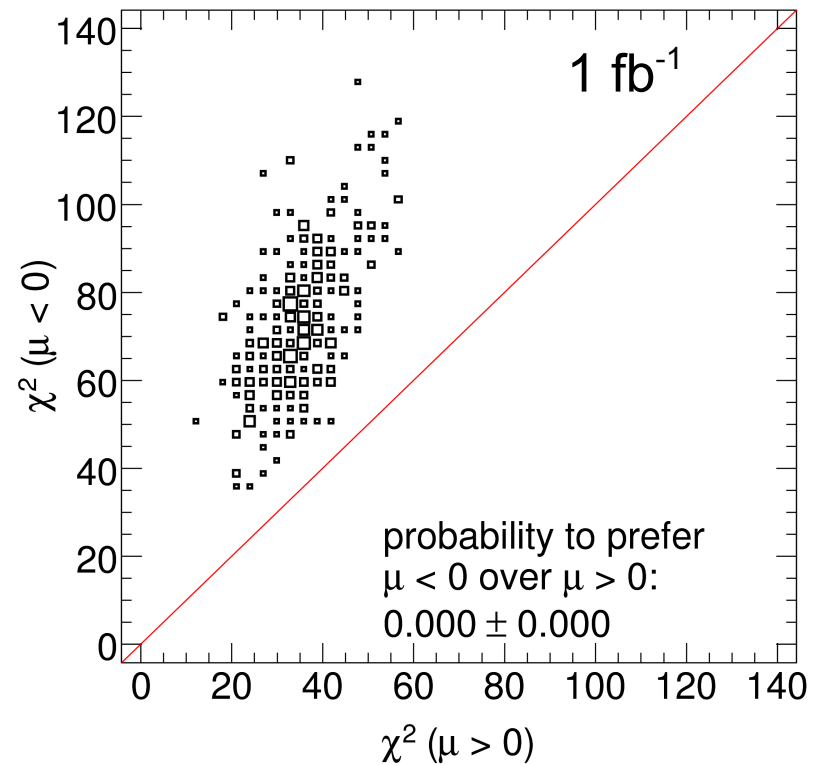
LE observables set to nominal SPS1a values for this combination

$\mu > 0$ vs. $\mu < 0$ from LHC+LE

LHC



LHC+LE



MSSM18 fit to LHC300+LE

So far we always assumed certain SUSY breaking mechanism

Can we also fit more general models to LHC+LE observables?

YES, WE CAN!

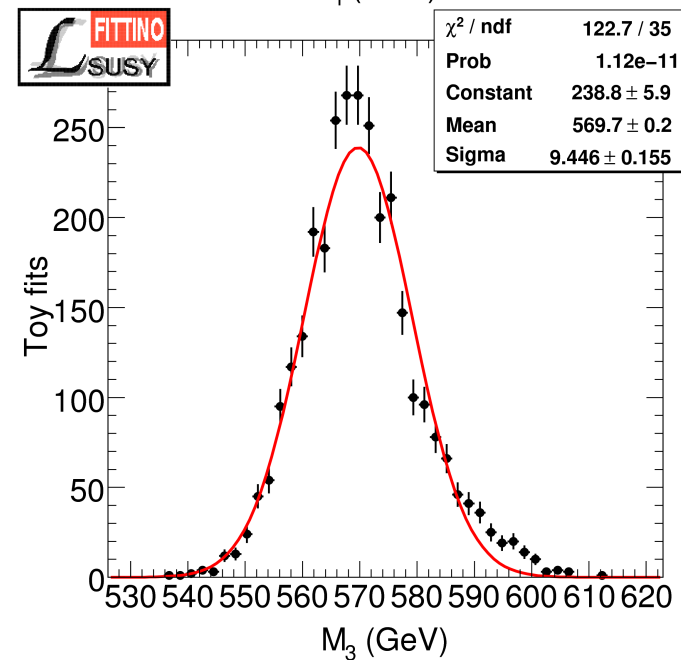
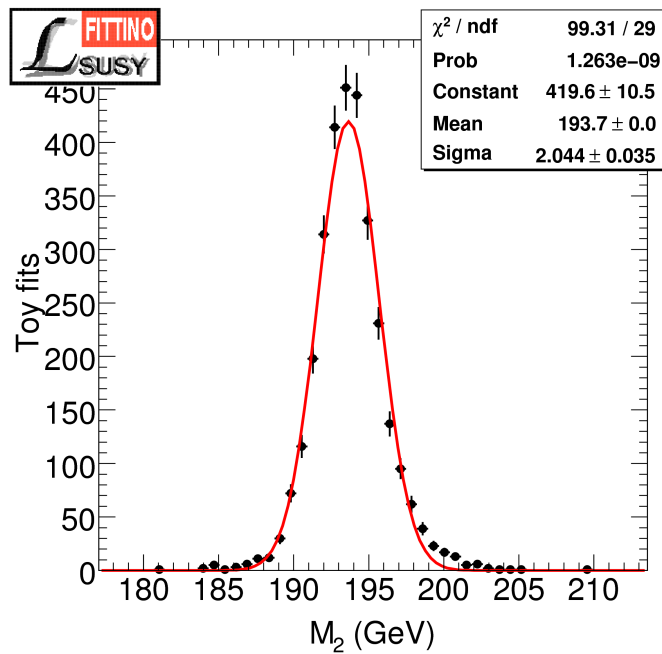
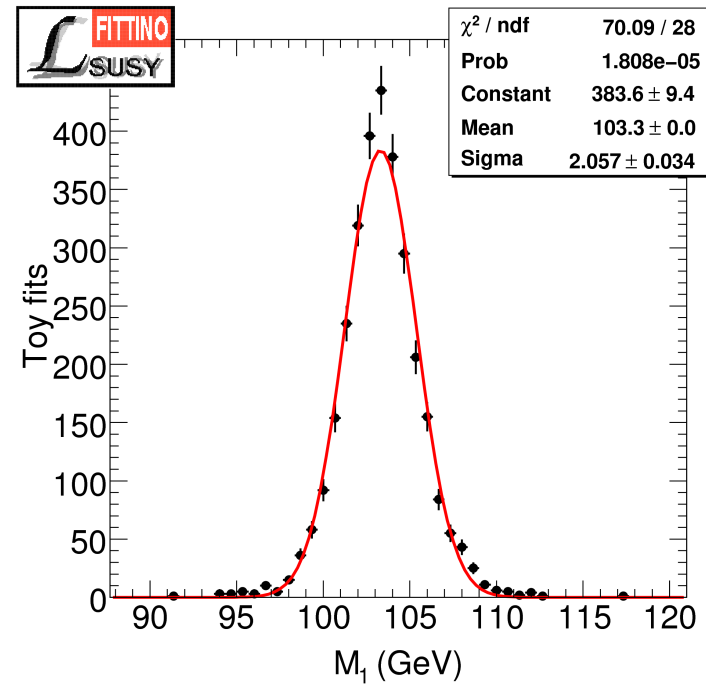
Assumptions:

- No CP violation (all phases = 0)
- No mixing between generations
- No mixing within the first two generations
- Universality of same-type sfermion mass parameters in first two generations

→ MSSM18

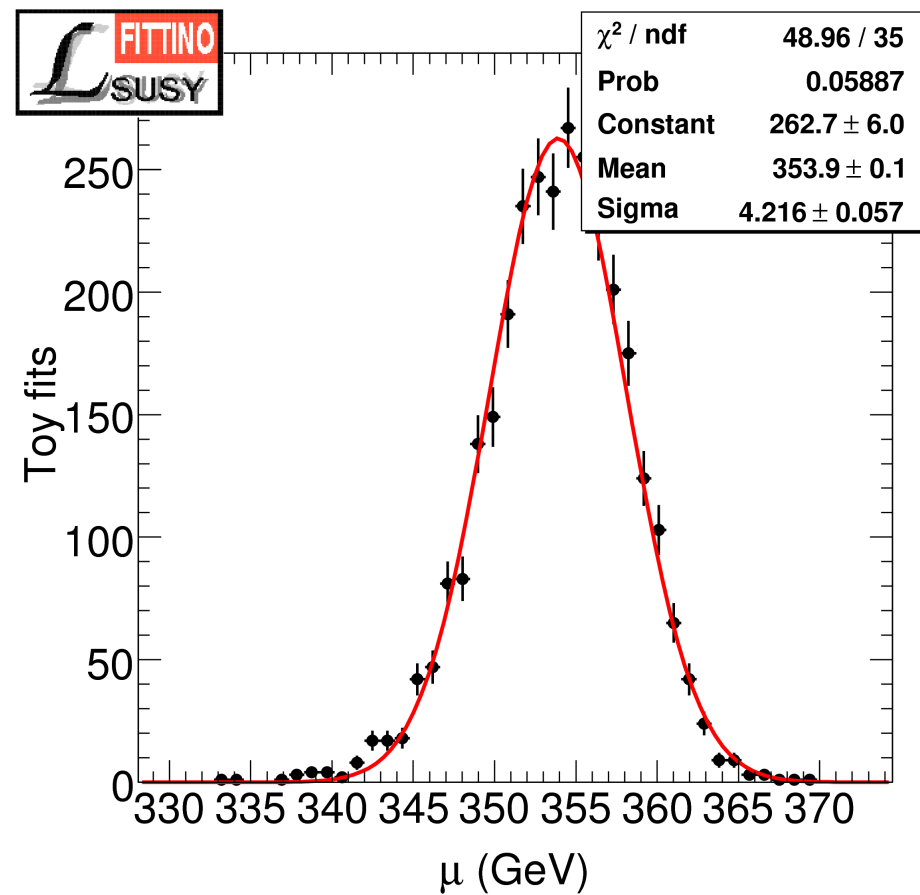
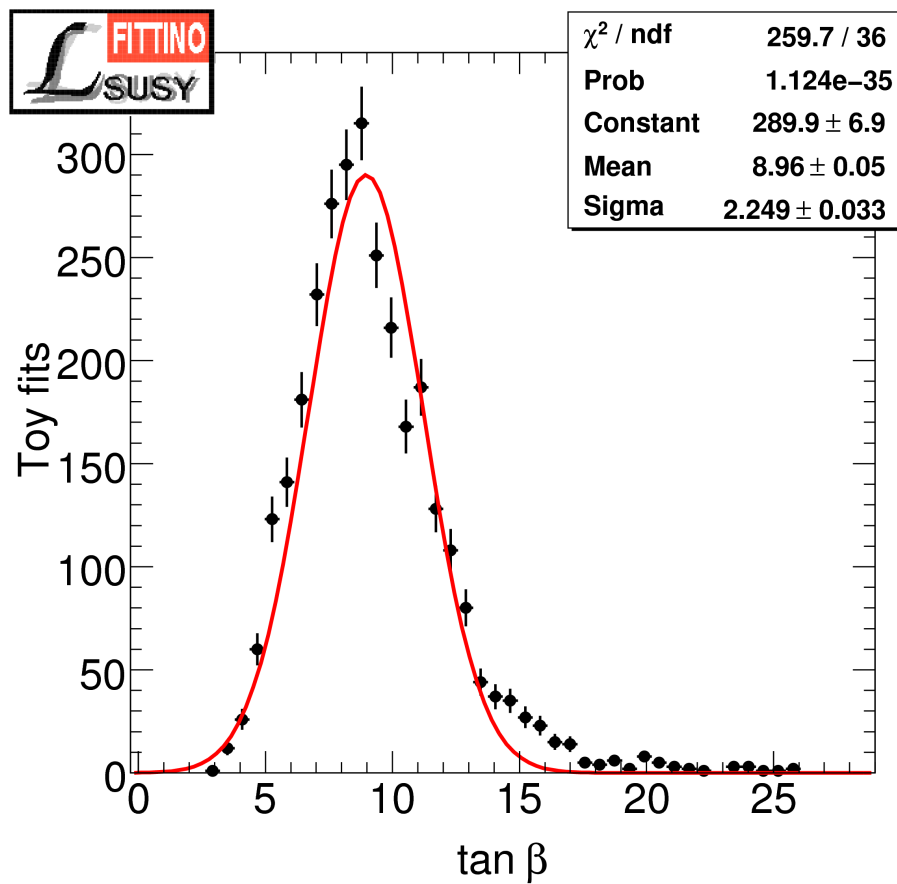
MSSM18 fit to LHC300+LE

Excerpts from toy fit parameter distributions...



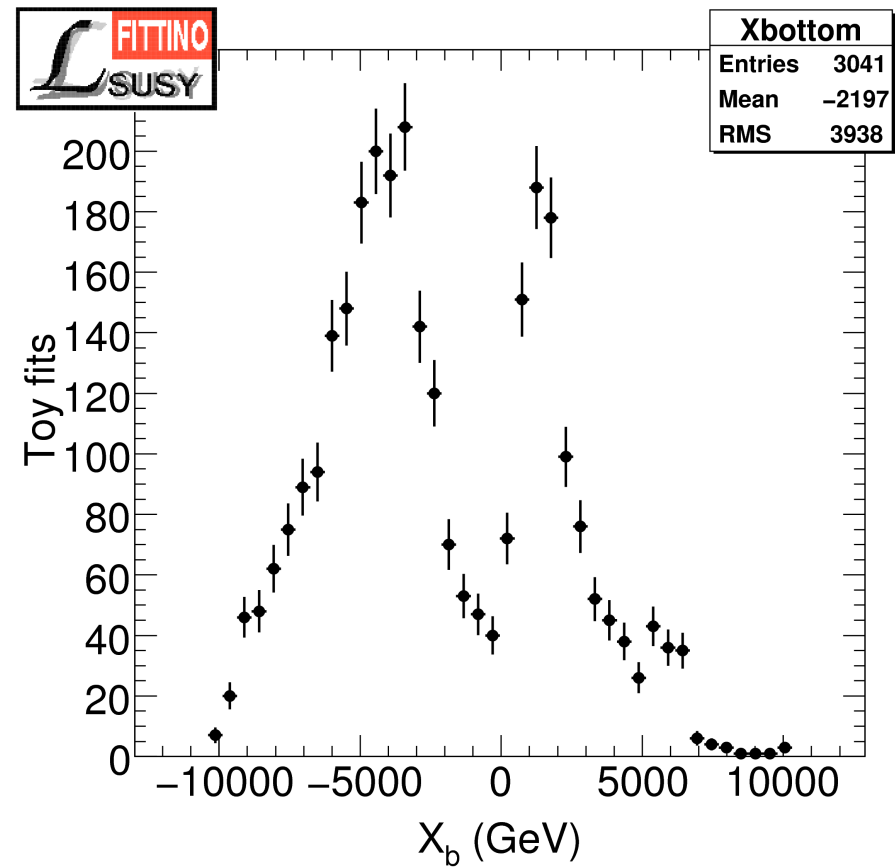
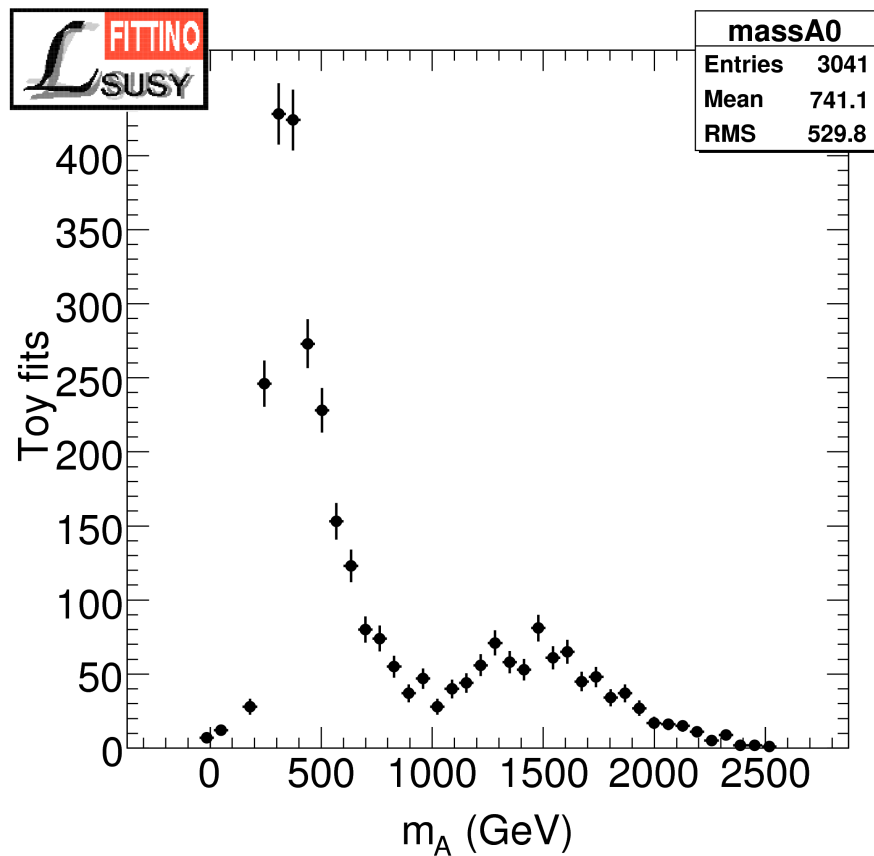
MSSM18 fit to LHC300+LE

Excerpts from toy fit parameter distributions...



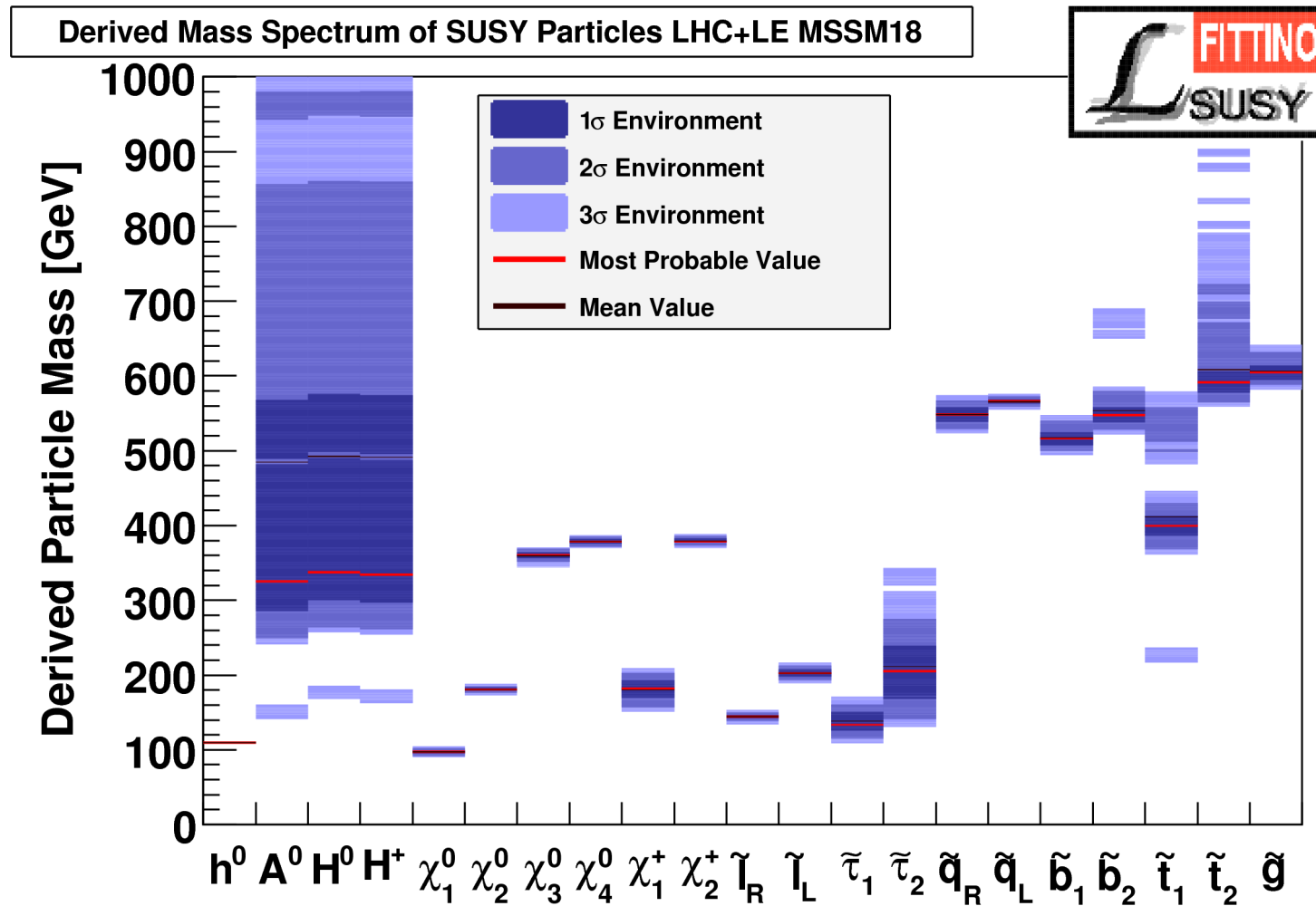
MSSM18 fit to LHC300+LE

Excerpts from toy fit parameter distributions...

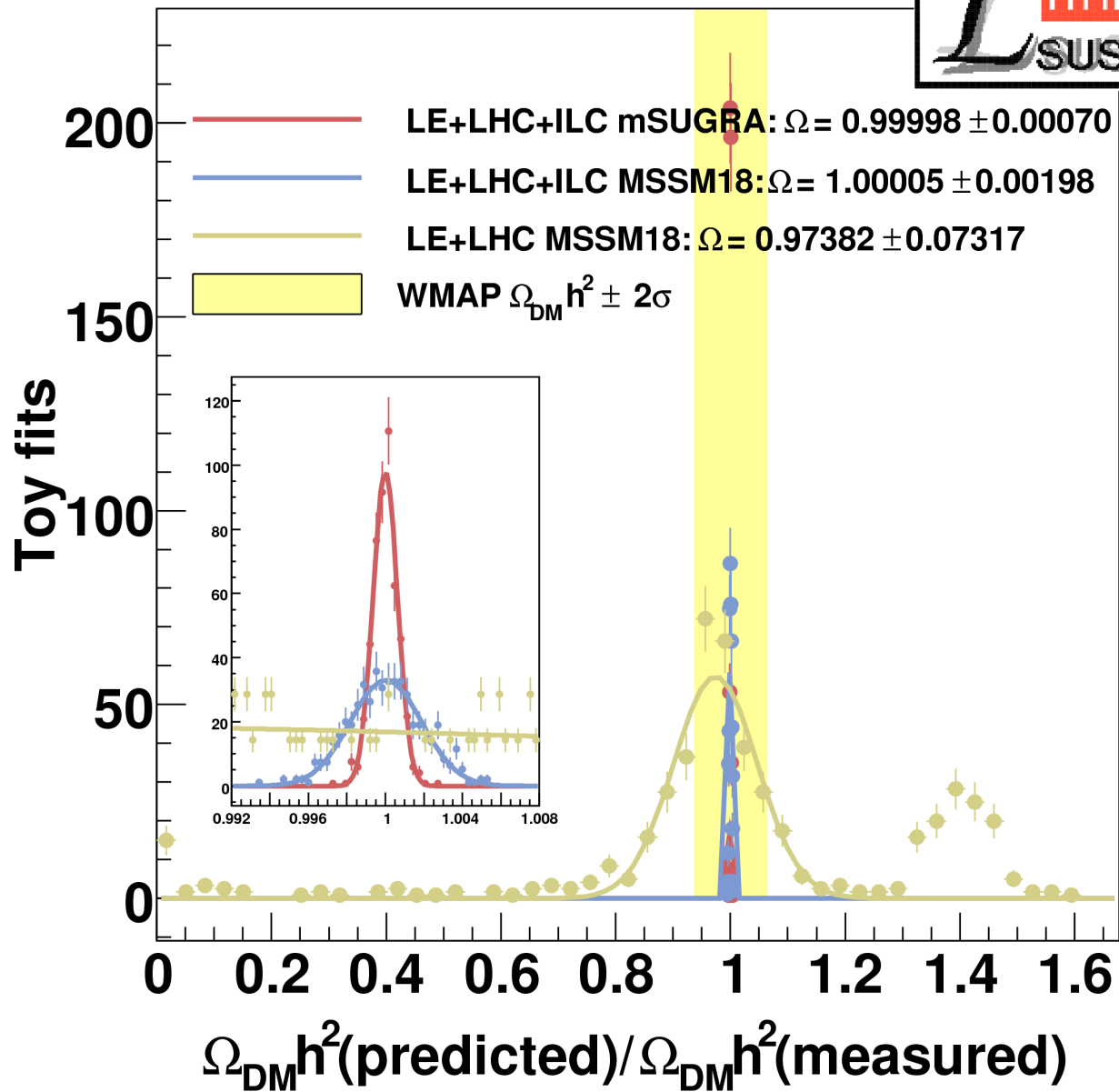


$$X_b = A_b - \mu \tan \beta$$

Mass spectrum from MSSM18 fit (LHC300+LE)



Relic density



Summary

- Discovery of new physics at the LHC **might** be the “easy” part (if Nature is not too nasty)
- Pinning down the underlying model **might** be more demanding
- LE measurements favour SUSY masses ≤ 1 TeV
- LE measurements **might** still provide valuable constraints for SUSY in the first phase of the LHC
- LHC results pretty powerful in constrained SUSY models
- LHC results **might** exhibit ambiguities in more general SUSY models \rightarrow ILC
- **Might, might, might** ... LHC will hopefully end speculations soon!