

Some Recent Developments in Dark Matter Theory

Leszek Roszkowski

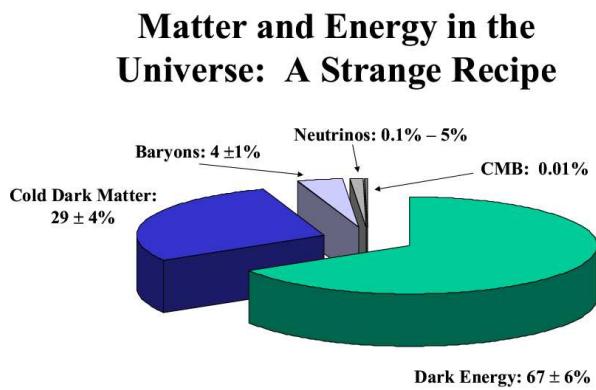
CERN

and

Astro–Particle Theory and Cosmology Group

Sheffield, England

Cosmic Inventory



- $\Omega_{\text{DM}} h^2 = 0.104 \pm 0.009$
 - $\Omega_b h^2 = 0.0224 \pm 0.0009$
- ⇒ most matter non–baryonic
WIMP: best candidate around

WIMPs - Several Directions and Tools

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- DM software packages (DarkSusy, Micromegas)
- related SUSY packages (Softsusy, Suspect, FeynHiggs, ...)
- new: Markov Chain Monte Carlo scan + probabilistic approach

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⇒ impact: WIMP DM studies has become a mainstream activity

supported by ILIAS/ENTApP

SUSY Models

Two basic approaches:

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- general MSSM

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 - SO(10)–GUT
 - Next-to-MSSM (NMSSM)
 - ...

The MSSM

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Minimal Supersymmetric

Standard Model

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gauge bosons

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WIMP: “neutralino” χ = lightest mass e’state of neutral gauginos and higgsinos

Constrained MSSM

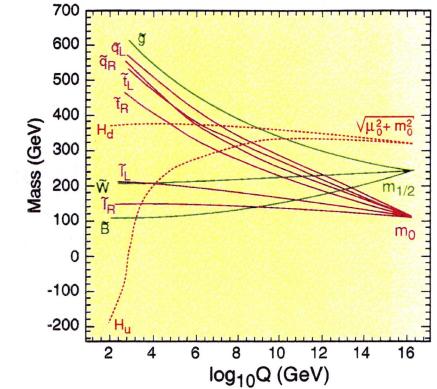
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At M_{GUT} :

- gauginos $M_1 = M_2 = m_{\tilde{g}} = m_{1/2}$ (c.f. MSSM)
- scalars $m_{\tilde{q}_i}^2 = m_{\tilde{l}_i}^2 = m_{H_b}^2 = m_{H_t}^2 = m_0^2$
- 3-linear soft terms $A_b = A_t = A_0$



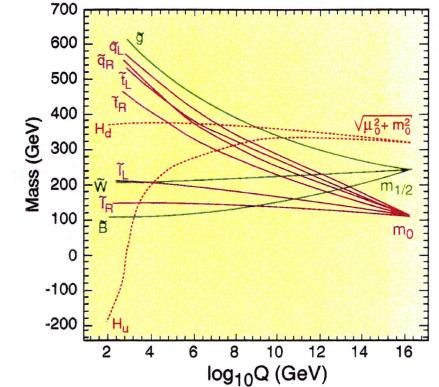
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- radiative EWSB

$$\mu^2 = \frac{(m_{H_b}^2 + \Sigma_b^{(1)}) - (m_{H_t}^2 + \Sigma_t^{(1)}) \tan^2 \beta}{\tan^2 \beta - 1} - \frac{m_Z^2}{2}$$

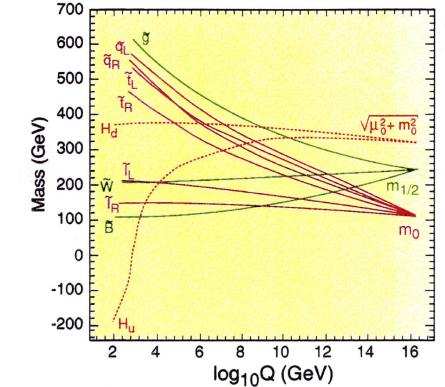


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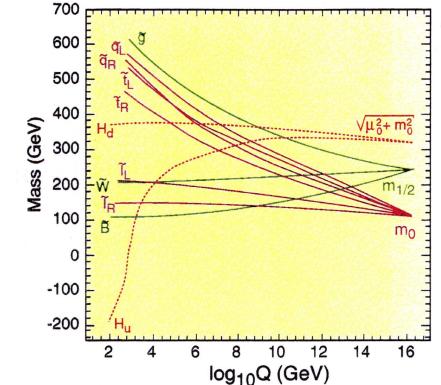
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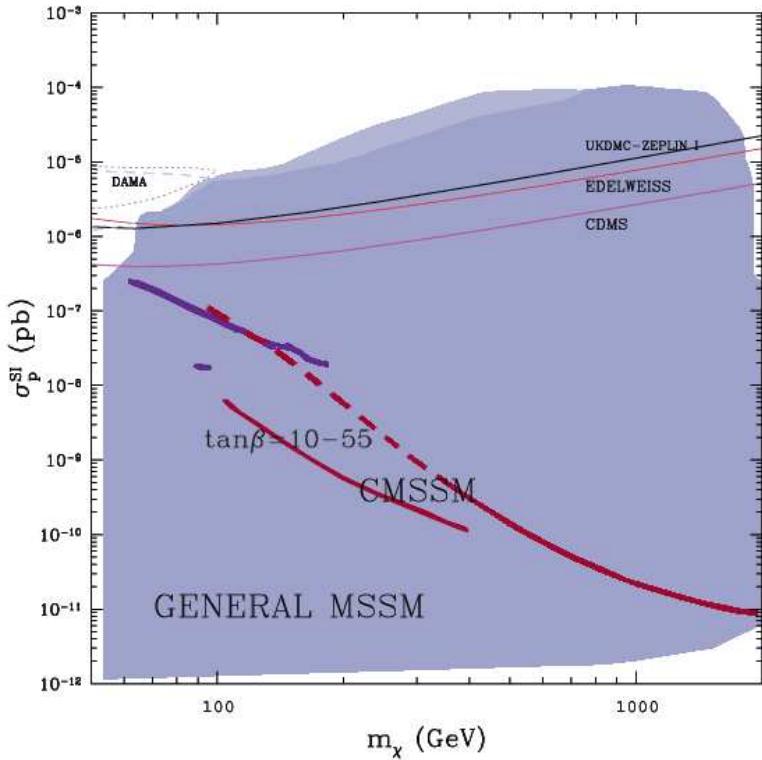
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- five independent parameters: $\tan \beta$, $m_{1/2}$, m_0 , A_0 , $\text{sgn}(\mu)$
- mass spectra at m_Z : run RGEs, 2-loop for g.c. and Y.c, 1-loop for masses
- some important quantities (μ , m_A , ...) very sensitive to procedure of computing EWSB & minimizing V_H

we use SoftSusy and FeynHiggs

Expectations for σ_p^{SI} with unification



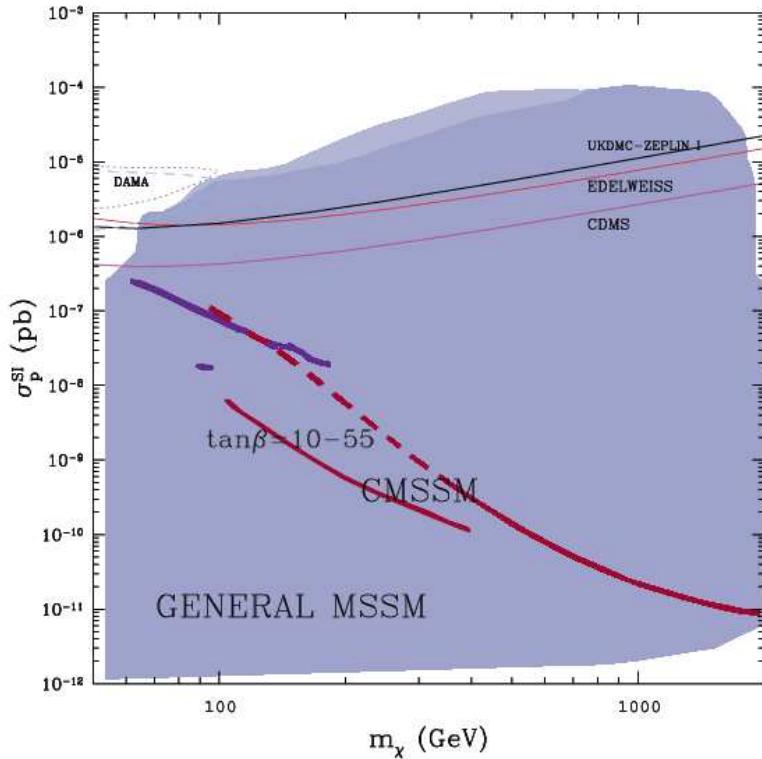
σ_p^{SI} – WIMP–proton SI elastic scatt. c.s.

blue: general MSSM

red: Constrained MSSM

$A_0 = 0, \tan\beta = 10, 55$

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- effect of varying CMSSM and SM inputs?

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- What is really meant by (dis-)allowed regions?
- nr of scan points $\propto k^N$ (where k : nr of steps, N : nr of parameters)
- \Rightarrow highly inefficient for $N \gtrsim$ a few
- hard to incorporate other uncertainties (SM, theory, ...), and assess their impact

MCMC + Bayesian Statistics

(MCMC=Markov Chain Monte Carlo)

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allows to make global statements, expose correlations, etc.

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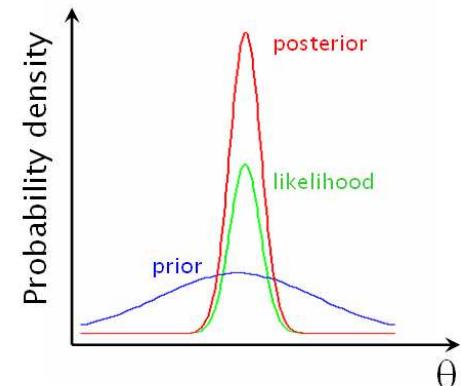
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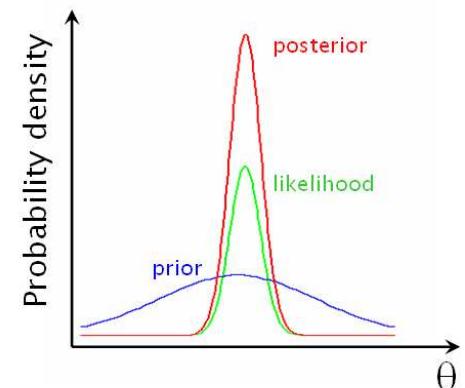
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$$p(\theta, \psi | d) = \frac{p(d|\xi)\pi(\theta, \psi)}{p(d)}$$

- $p(d|\xi)$: likelihood
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- $p(d)$: evidence (normalization factor)



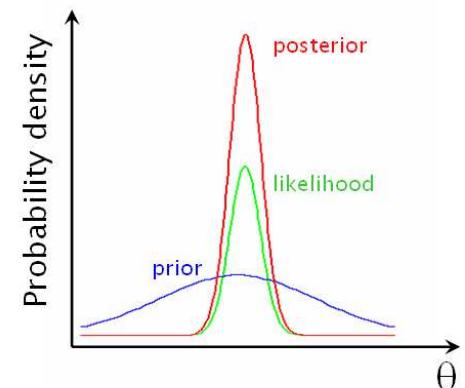
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- usually marginalize over SM (nuisance) parameters ψ : $p(\theta|d)$



Bayesian Analysis of the CMSSM

Roszkowski, Ruiz de Austri, Trotta, hep-ph/0602028, hep-ph/0611173

- $\theta = (m_0, m_{1/2}, A_0, \tan \beta)$: main CMSSM parameters (set $\mu > 0$)
- $\psi = (m_t^{\text{pole}}, m_b(m_b)^{\overline{MS}}, \alpha_{\text{em}}(M_Z), \alpha_s)$: SM (nuisance) parameters
- priors – assume flat distributions and ranges as:

CMSSM parameters θ	
“2 TeV range”	“4 TeV range”
$50 \text{ GeV} < m_0 < 2 \text{ TeV}$	$50 \text{ GeV} < m_0 < 4 \text{ TeV}$
$50 \text{ GeV} < m_{1/2} < 2 \text{ TeV}$	$50 \text{ GeV} < m_{1/2} < 4 \text{ TeV}$
$ A_0 < 5 \text{ TeV}$	$ A_0 < 7 \text{ TeV}$
$2 < \tan \beta < 62$	
flat priors: SM (nuisance) parameters ψ	
$160 \text{ GeV} < m_t^{\text{pole}} < 190 \text{ GeV}$	
$4 \text{ GeV} < m_b(m_b)^{\overline{MS}} < 5 \text{ GeV}$	
$0.10 < \alpha_s < 0.13$	
$127.5 < 1/\alpha_{\text{em}}(M_Z) < 128.5$	

Experimental Measurements

(assume Gaussian distributions)

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SM (nuisance) parameter	Mean	Error
	μ	σ (expt)
m_t^{pole}	171.4 GeV	2.1 GeV
$m_b(m_b)^{\overline{MS}}$	4.20 GeV	0.07 GeV
α_s	0.1176	0.002
$1/\alpha_{\text{em}}(M_Z)$	127.918	0.018

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new $M_W = 80.413 \pm 0.048$ GeV
 (Jan 07, not included)
 $\text{BR}(\bar{B} \rightarrow X_s \gamma) \times 10^4$:
 new SM: 3.15 ± 0.21 (Misiak &
 Steinhauser, Sept 06) used here
 2.98 ± 0.26 (Becher & Neubert,
 Nov 06)

Derived observable	Mean μ	Errors	
		σ (expt)	τ (th)
M_W	80.392 GeV	29 MeV	15 MeV
$\sin^2 \theta_{\text{eff}}$	0.23153	16×10^{-5}	15×10^{-5}
$\delta a_\mu^{\text{SUSY}} \times 10^{10}$	28	8.1	1
$\text{BR}(\bar{B} \rightarrow X_s \gamma) \times 10^4$	3.55	0.26	0.21
ΔM_{B_s}	17.33	0.12	4.8
$\Omega_\chi h^2$	0.119	0.009	$0.1 \Omega_\chi h^2$

take as precisely known: $M_Z = 91.1876(21)$ GeV, $G_F = 1.16637(1) \times 10^{-5}$ GeV $^{-2}$

Experimental Limits

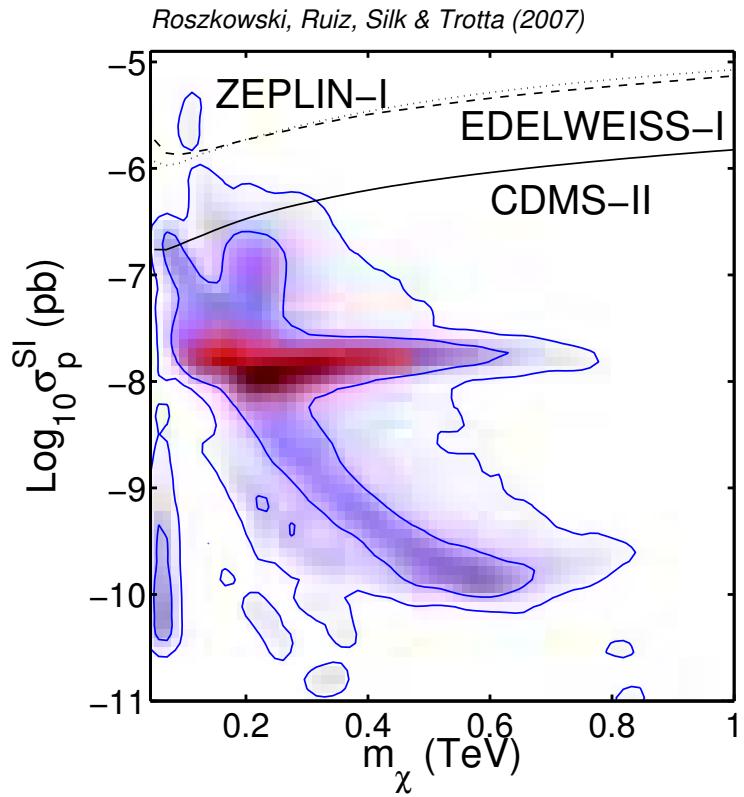
Derived observable	upper/lower limit	ξ_{lim}	Constraints
			τ (theor.)
$\text{BR}(\text{B}_s \rightarrow \mu^+ \mu^-)$	UL	1.5×10^{-7}	14%
m_h	LL	114.4 GeV (91.0 GeV)	3 GeV
$\zeta_h^2 \equiv g_{ZZh}^2 / g_{ZZH_{\text{SM}}}^2$	UL	$f(m_h)$	3%
m_χ	LL	50 GeV	5%
$m_{\chi_1^\pm}$	LL	103.5 GeV (92.4 GeV)	5%
$m_{\tilde{e}_R}$	LL	100 GeV (73 GeV)	5%
$m_{\tilde{\mu}_R}$	LL	95 GeV (73 GeV)	5%
$m_{\tilde{\tau}_1}$	LL	87 GeV (73 GeV)	5%
$m_{\tilde{\nu}}$	LL	94 GeV (43 GeV)	5%
$m_{\tilde{t}_1}$	LL	95 GeV (65 GeV)	5%
$m_{\tilde{b}_1}$	LL	95 GeV (59 GeV)	5%
$m_{\tilde{q}}$	LL	318 GeV	5%
$m_{\tilde{g}}$	LL	233 GeV	5%
(σ_p^{SI})	UL	WIMP mass dependent	$\sim 100\%$)

Note: DM direct detection σ_p^{SI} not applied due to astroph'l uncertainties (eg, local DM density)

Implications for direct detection: σ_p^{SI}

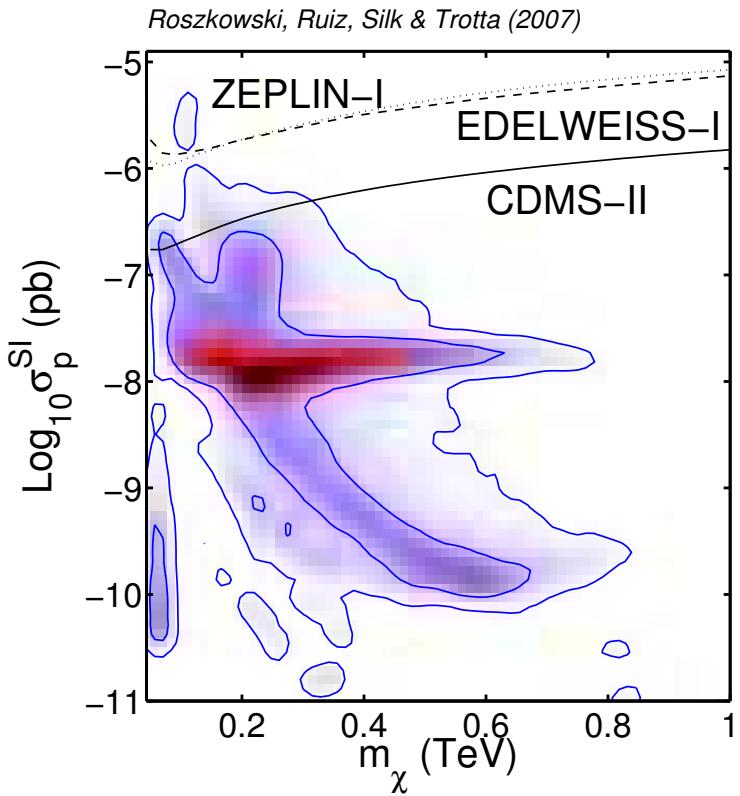
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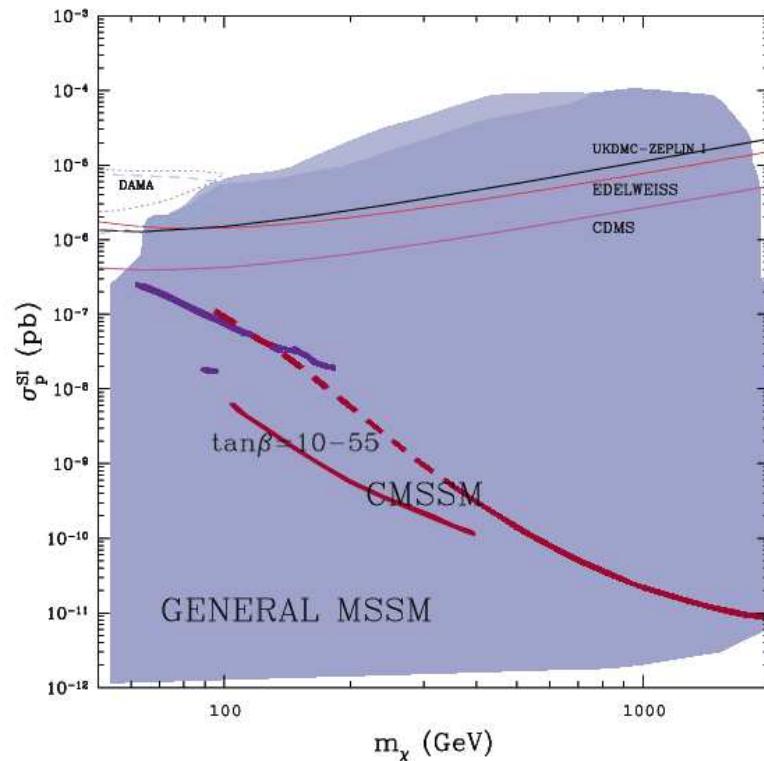


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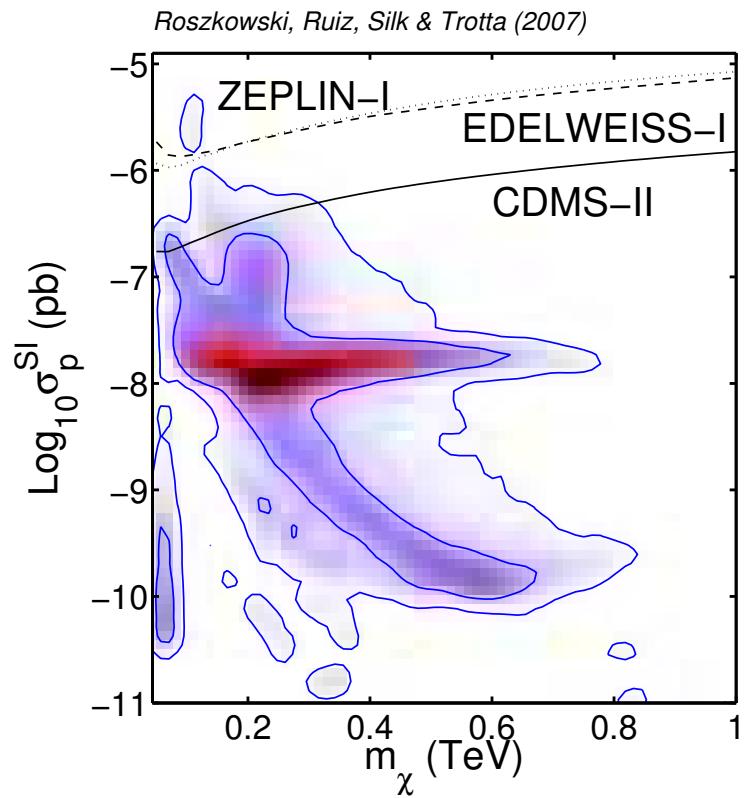


compare: fixed grid scan



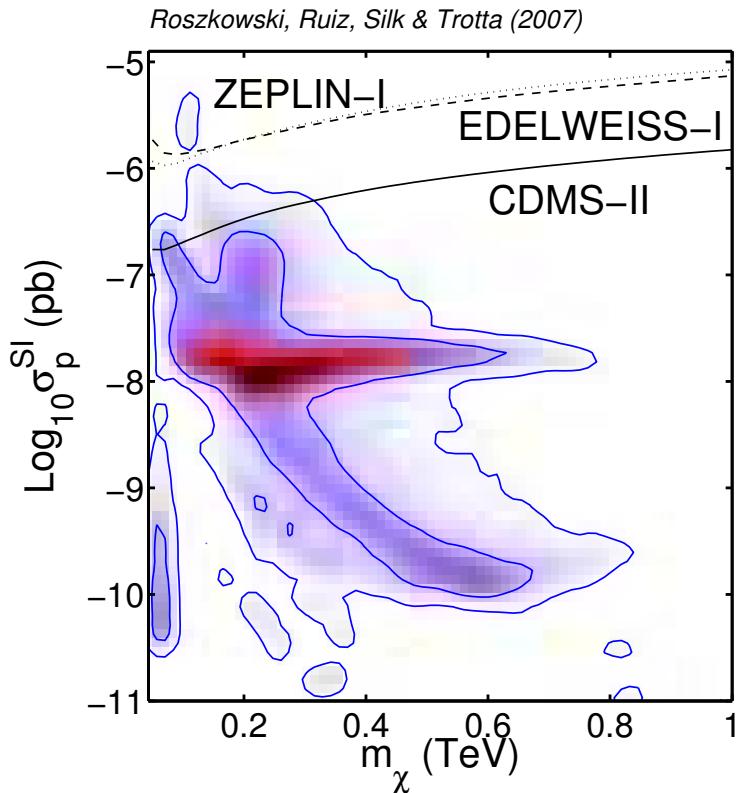
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CDMS (this year?):

will reach down to $\sigma_p^{SI} \sim 10^{-8}$ pb:
also Edelweiss-II (?)

⇒ explore the FP region

(large $m_0 \gg m_{1/2}$), outside of the LHC
reach

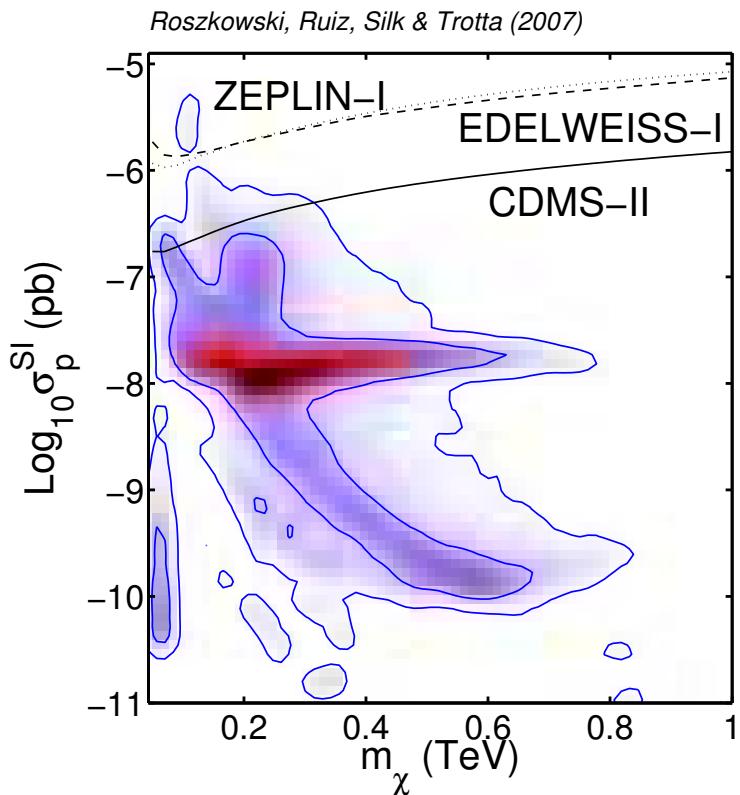
ultimately: “1 tonne” detectors:

$\sigma_p^{SI} \lesssim 10^{-10}$ pb

will cover all 68% region

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most probable range: 10^{-8} pb $\lesssim \sigma_p^{SI} \lesssim 10^{-10}$ pb

partly outside of the LHC reach ($m_\chi \lesssim 400$ GeV)

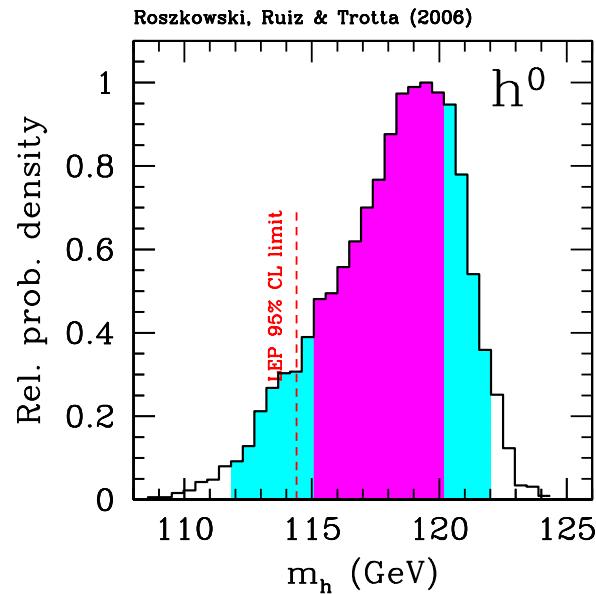
CMSSM Higgs Boson & the Tevatron

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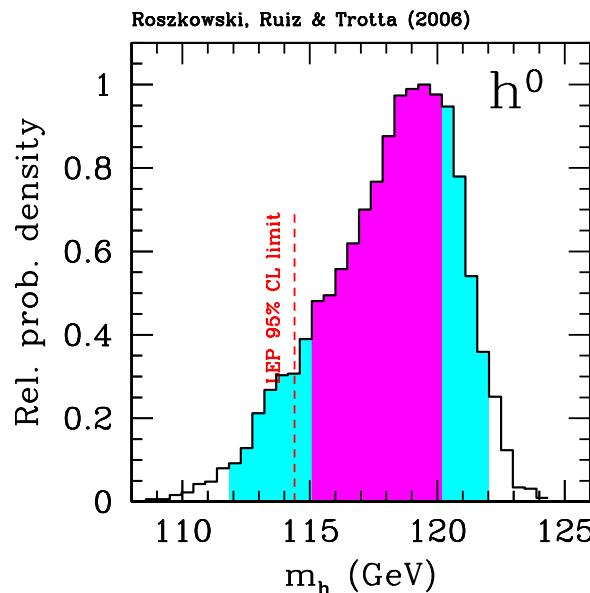
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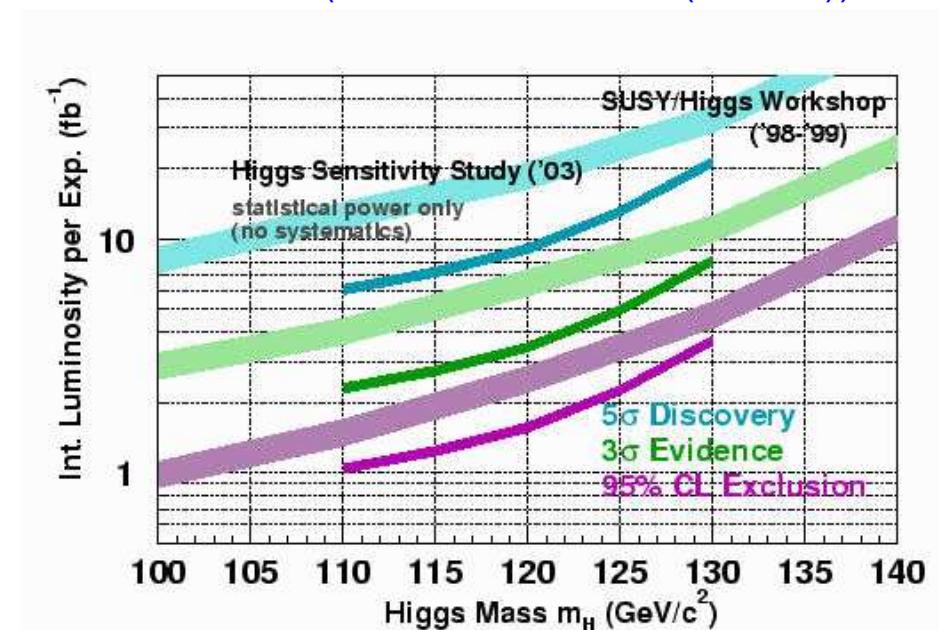
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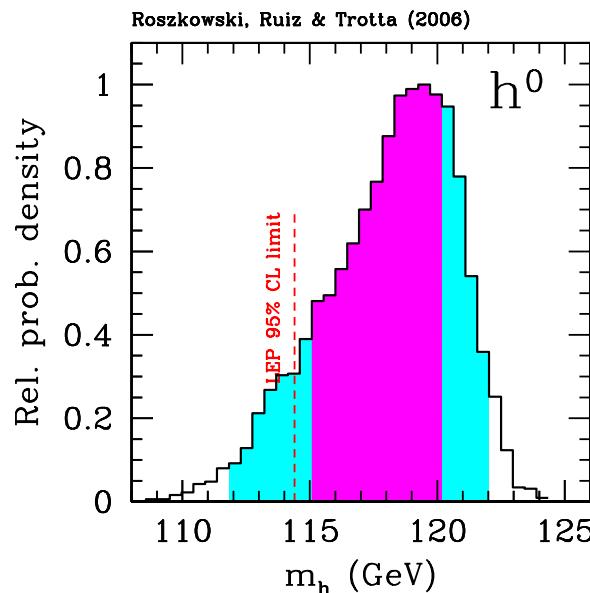
Tevatron reach (CDF and D0 WG (Oct 03))



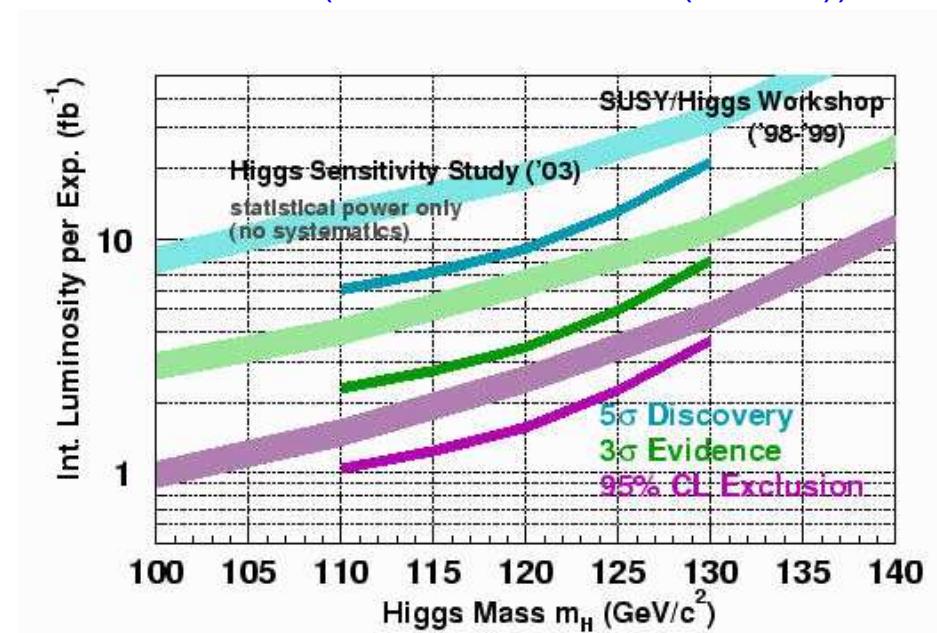
CMSSM Higgs Boson & the Tevatron

CMSSM: light Higgs boson h^0 is SM-like (SM-like couplings)

MCMC scan, Bayesian analysis



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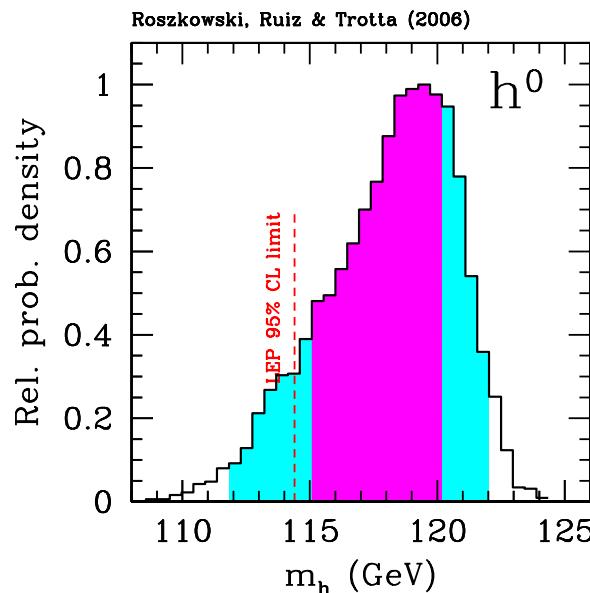
2 fb^{-1} /experiment already on tape

\Rightarrow enough to set 95% CL exclusion limit on 95% range of m_h

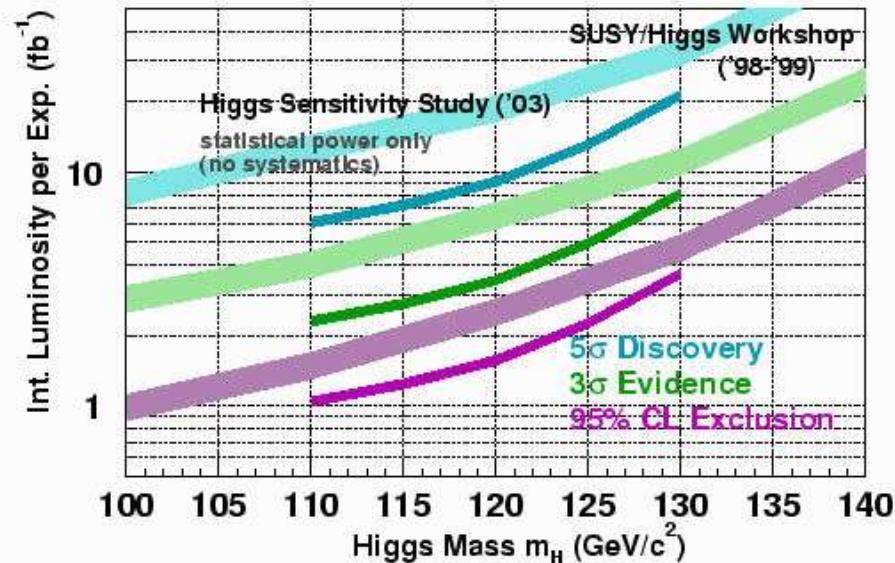
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2 fb^{-1} /experiment already on tape

⇒ enough to set 95% CL exclusion limit on 95% range of m_h

...or else...

with 4 fb^{-1} /expt: 3 σ evidence over entire 95% range of m_h

with $\sim 10 - 12 \text{ fb}^{-1}$ /expt: 5 σ discovery over entire 95% range of m_h

Tevatron: hope for up to $\sim 8 \text{ fb}^{-1}$ /expt

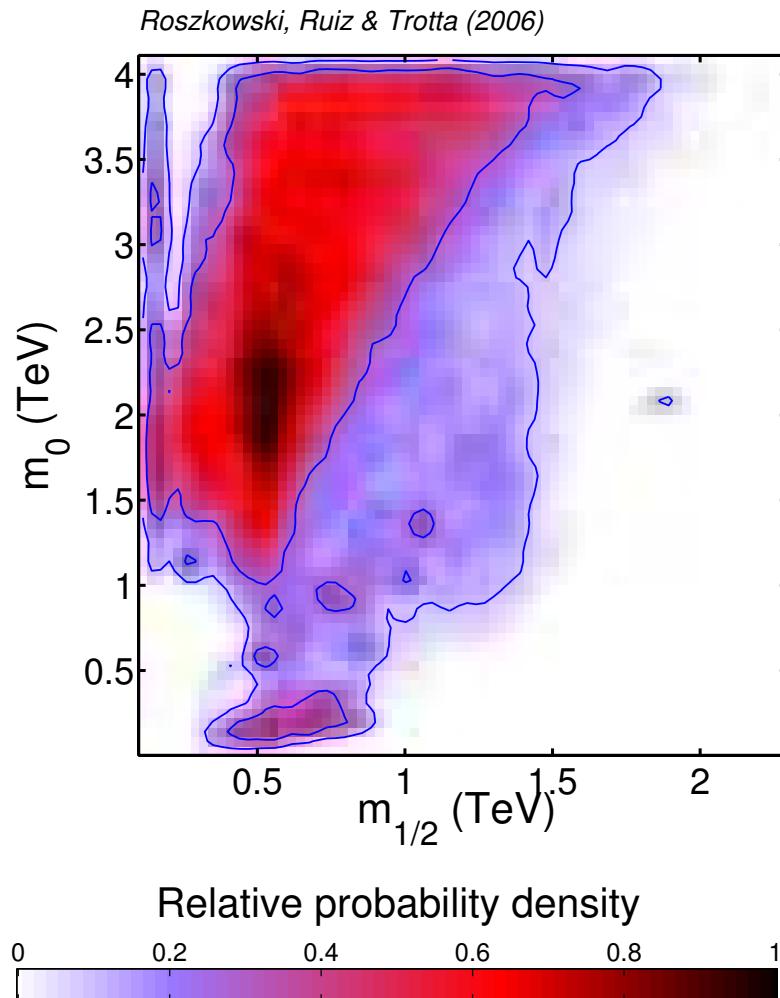
Full MCMC Scan of the CMSSM

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RRT update, in prep

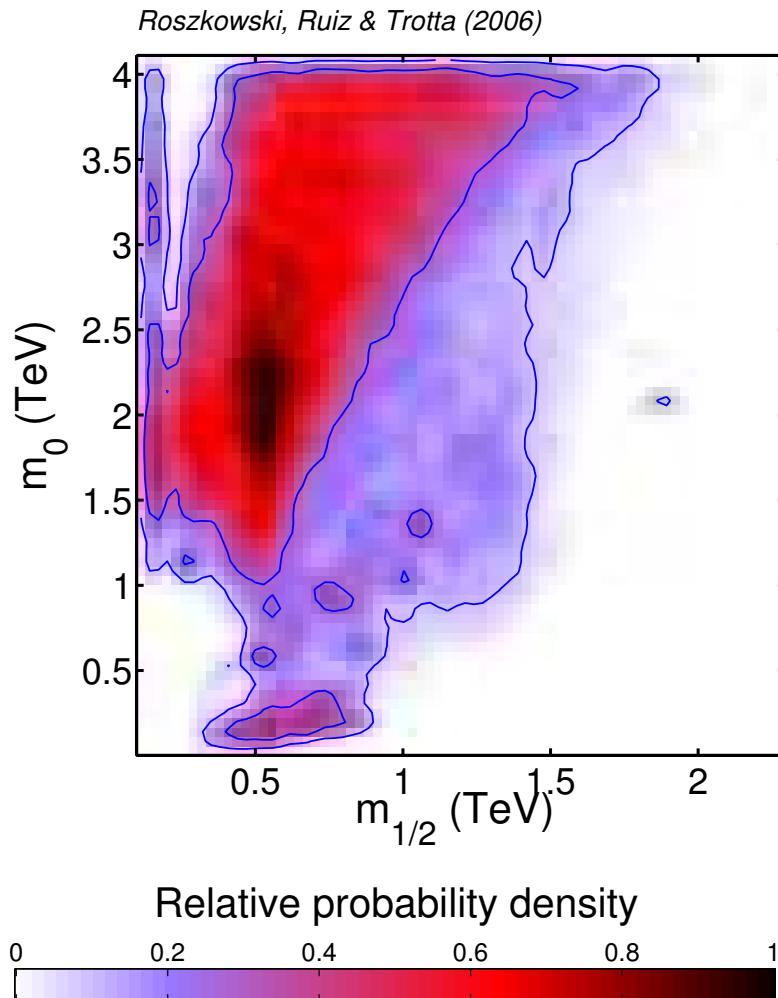


Bayesian analysis
relative probability density fn
flat priors
 $p(m_0, m_{1/2} | d)$

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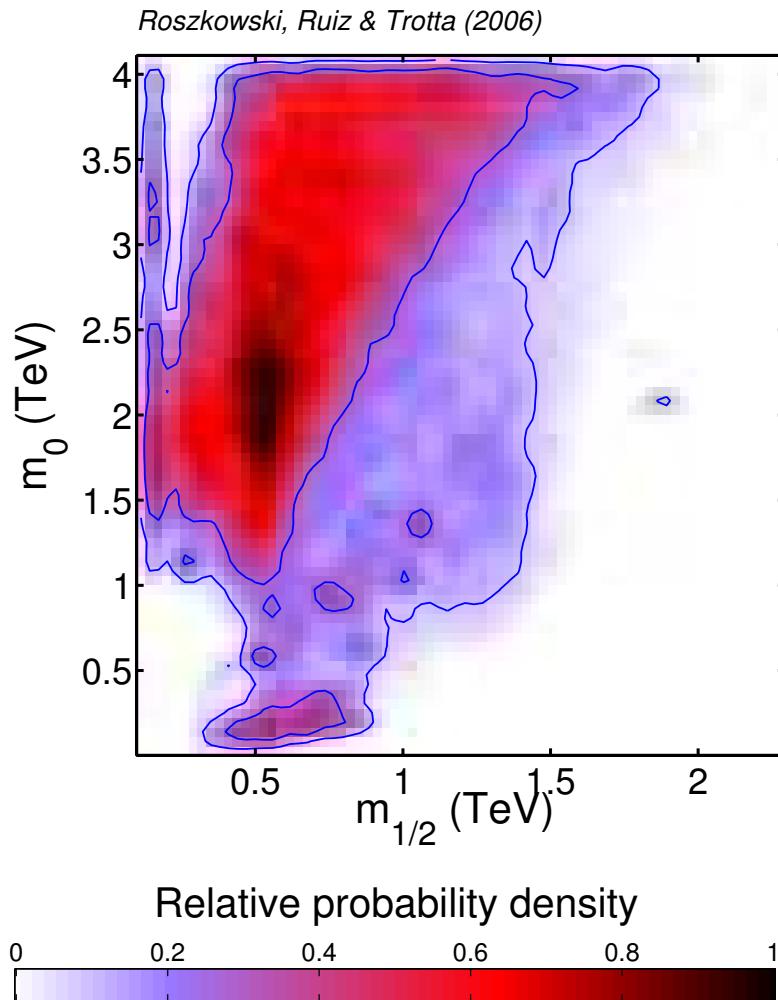
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similar study by Allanach+Lester (no mean qof),
see also, Ellis et al (EHOW, χ^2 approach, no MCMC)

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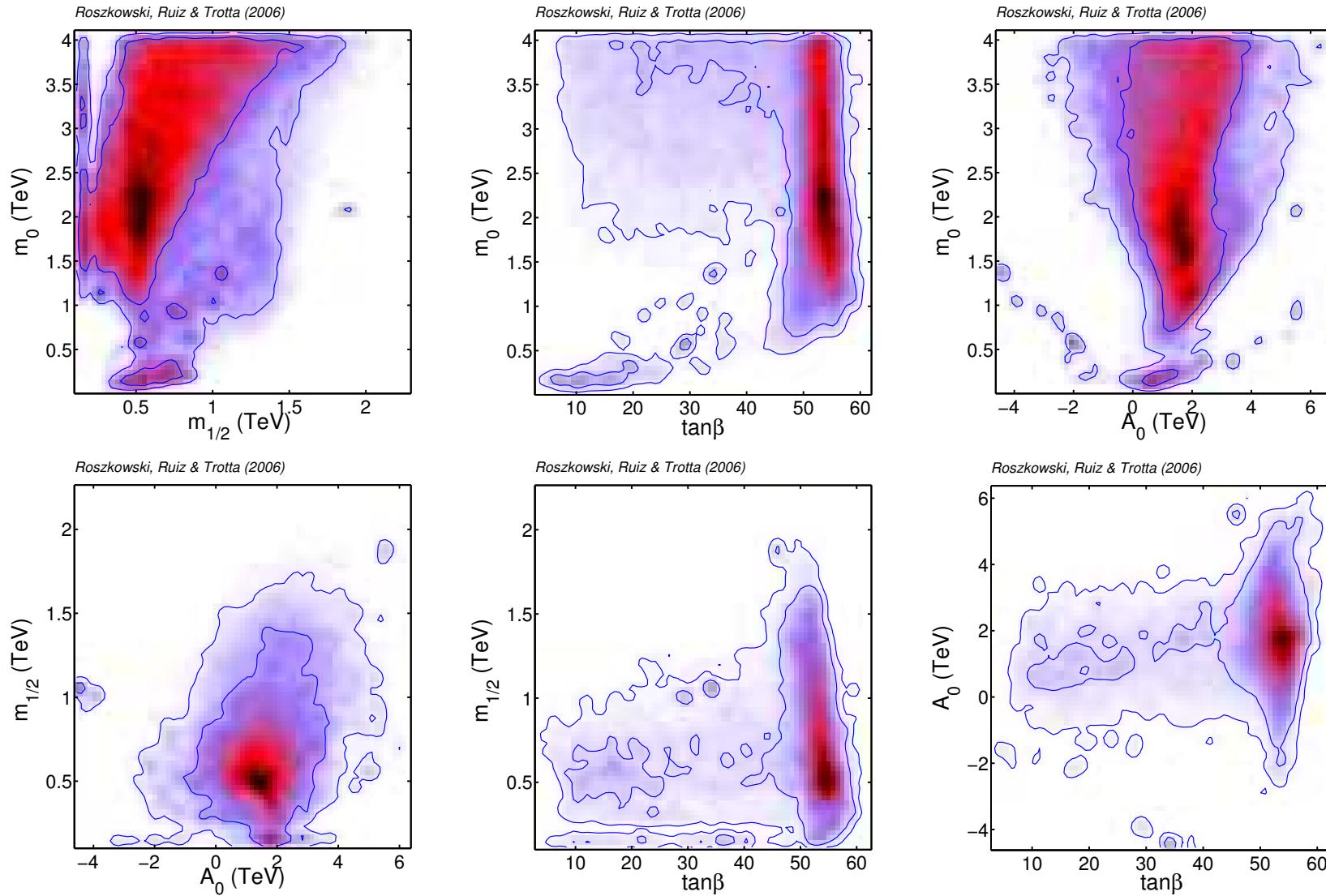
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unlike others (except for A+L), we vary also SM parameters

Full MCMC Scan of the CMSSM



peak at $A_0 \simeq 1.5$ TeV (in grid scans one usually sets $A_0 = 0$)

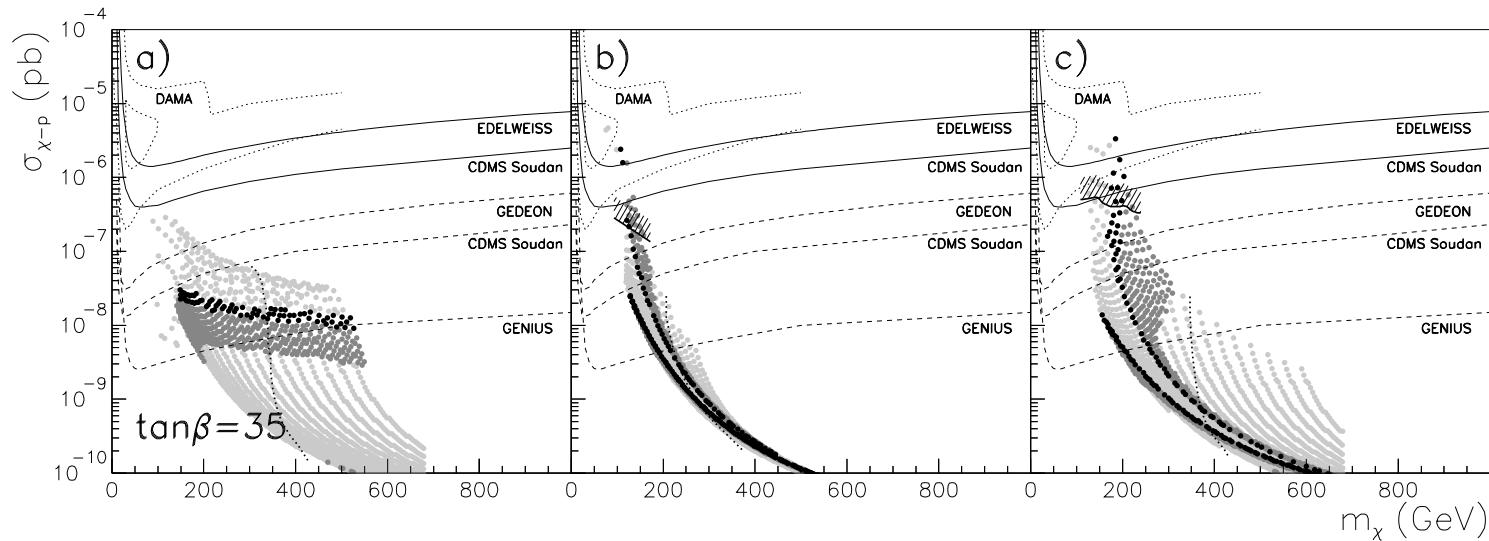
WIMP detection in NUHM models

NUHM = non-unified Higgs mass

at M_{GUT} : $m_{H_b}^2 = (1 + \delta_b)m_0^2$, $m_{H_t}^2 = (1 + \delta_t)m_0^2$

implications for σ_p^{SI}

left: $\delta_b = 0, \delta_t = 1$, center: $\delta_b = -1, \delta_t = 0$, right: $\delta_b = -1, \delta_t = 1$



Baek, et al, hep-ph/0505019

new SM value for $b \rightarrow s\gamma$ not included

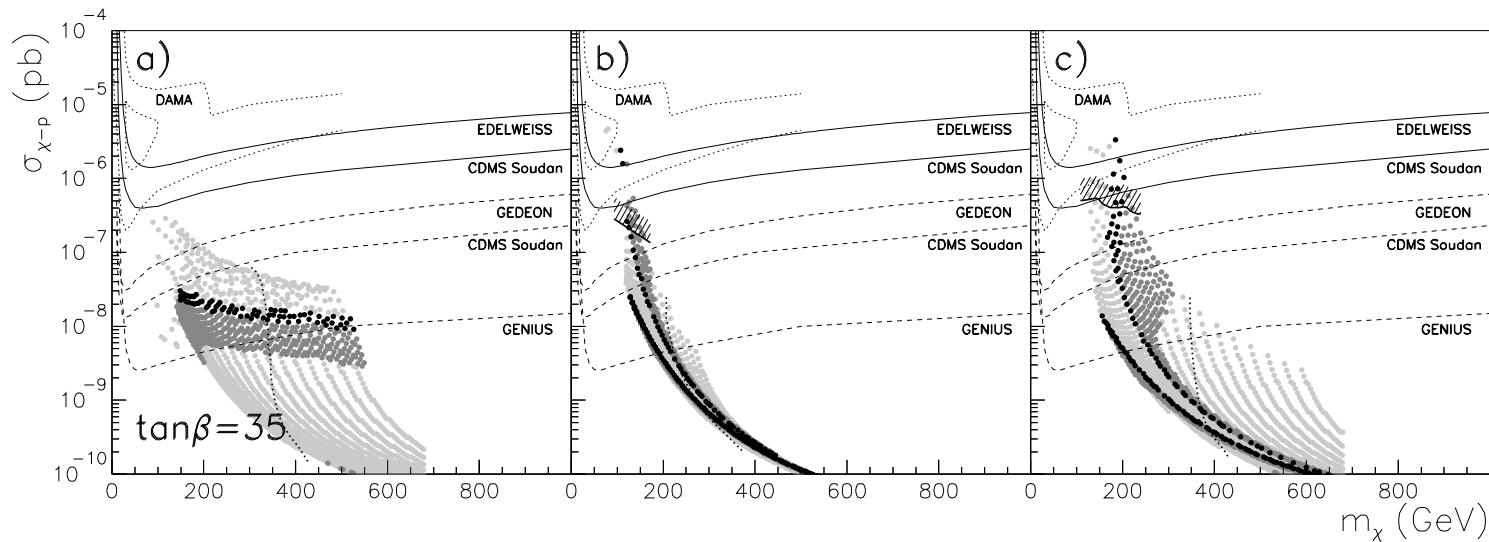
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⇒ for $\delta_b < 0$ and $\delta_t > 0$, σ_p^{SI} can be larger than in the CMSSM

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- it would be helpful if SUSY were found at LHC...

Impact of $b \rightarrow s\gamma$

recall

$$BR(B \rightarrow X_s \gamma) = B(W^-/t) + B(H^-/t) - \text{sgn}(\mu) B(\chi^-/\tilde{t})$$

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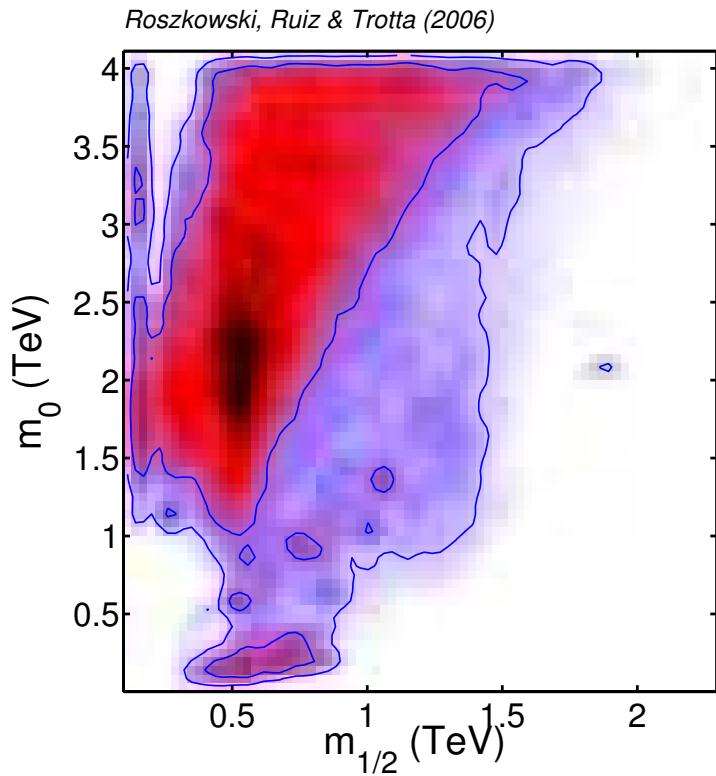
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EXPT: 3.55 ± 0.26 , TH: 3.15 ± 0.21

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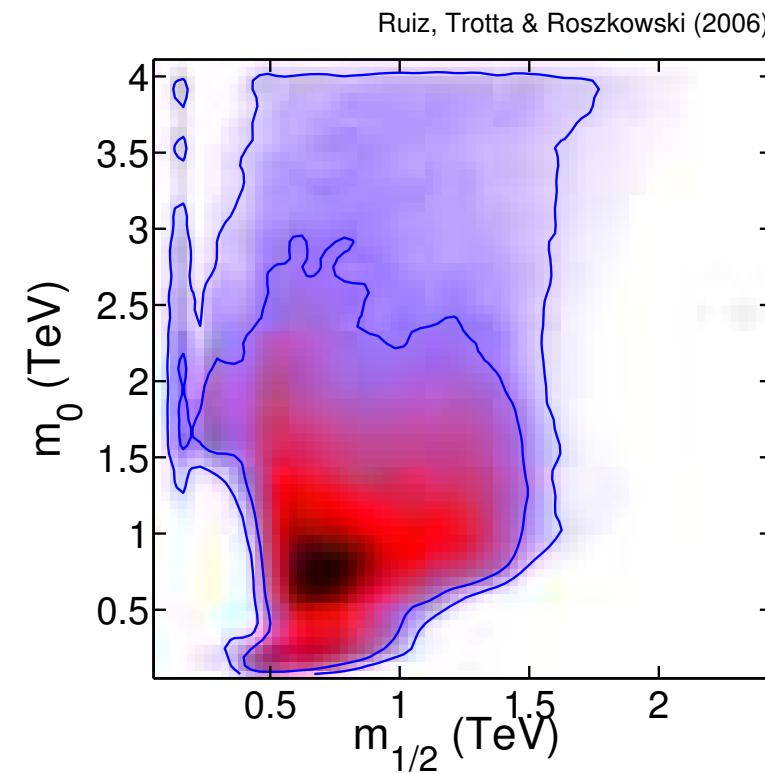
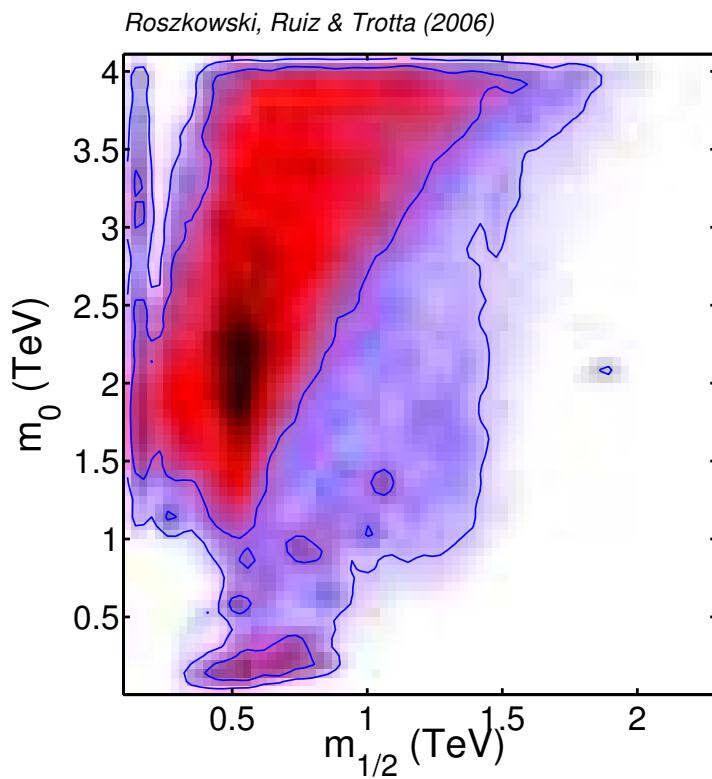
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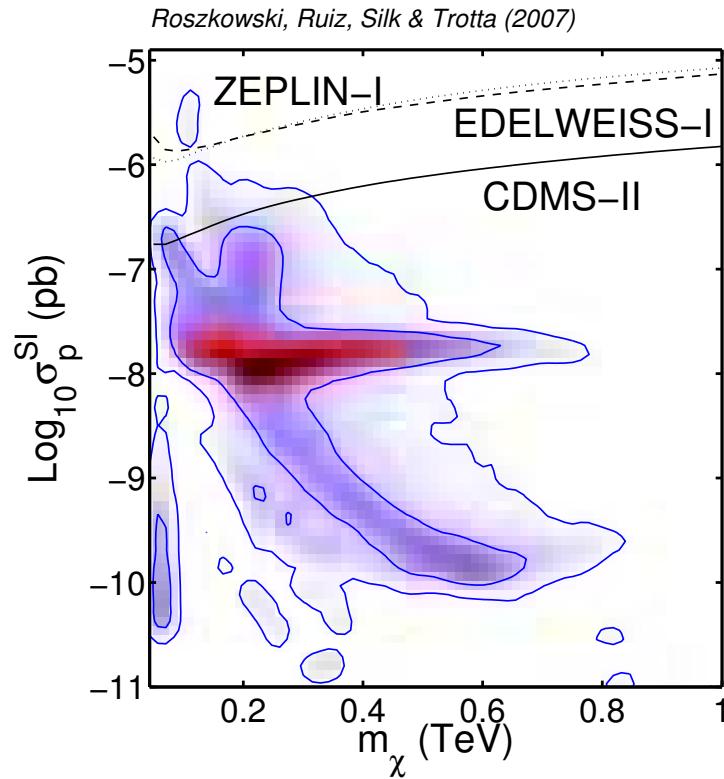
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