

Measuring Higgs Couplings

Tilman Plehn

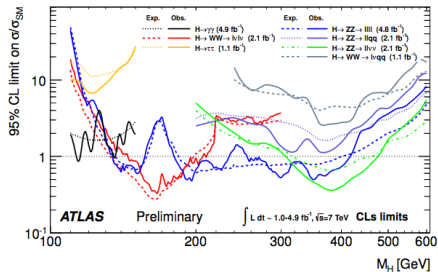
Universität Heidelberg

Pheno 2012, Pittsburgh

Data!

Around Moriond 2012

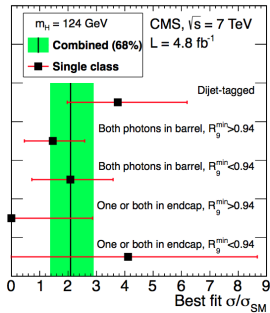
- ATLAS and CMS results published
- official line: ‘exclusion gone wrong’ [slowly changing]



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- mass and rate from $H \rightarrow \gamma\gamma$



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Likelihoods

Parameters

Future

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- ATLAS and CMS results published
 - official line: ‘exclusion gone wrong’ [slowly changing]
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 - mass and rate from $H \rightarrow \gamma\gamma$
 - too early for model building
 - what about (standard) model testing?
- ⇒ **establish technique**

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Why 125 GeV is just perfect [Zeppenfeld et al; Dührssen et al; SFitter 2009; SFitter-Higgs soon]

- many parameters: Higgs couplings to $W, Z, t, b, \tau + (g, \gamma)$ [SM-like operators]

$$g_{Hxx} \equiv g_x = g_x^{\text{SM}} (1 + \Delta_x)$$

- many measurements: $GF : H \rightarrow ZZ, WW, \gamma\gamma$ [already 2011]
- $WBF : H \rightarrow ZZ, WW, \gamma\gamma, \tau\tau$ [mostly 2012]
- $VH : H \rightarrow b\bar{b}$ [2014]
- $t\bar{t}H : H \rightarrow \gamma\gamma, b\bar{b}$ [2014]

⇒ perfect application for SFitter

Global: likelihood maps

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Markov chain probability map [SFitter]

- LHC parameters: weak-scale Lagrangean [Higgs, MSSM, dark matter,...]
- errors: correlated Poisson-Gaussian-RFit
- likelihood map: data given a model $p(d|m) \sim |\mathcal{M}|^2(m)$

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Global view of 2011 data [Klute, Lafaye, TP, Rauch, Zerwas, Dührssen]

- is there a SM-like solution?
are there alternative solutions?

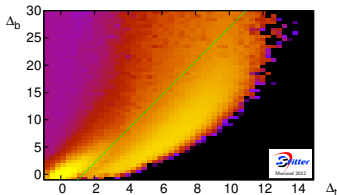
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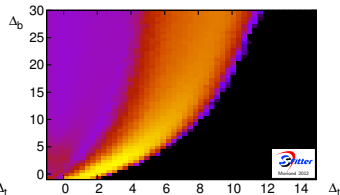
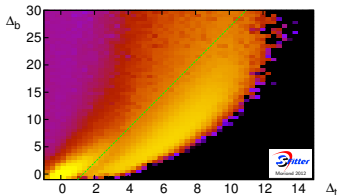
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- (1) expected 2011 results: SM central values, measured error bars
 - large-coupling solution separable
 - (2) measured 2011 results: measured central values and error bars
 - both solutions overlapping



Local: parameter extraction

Parameters with error bars [SFitter]

- starting with best MC point
- best fit point from Minuit
- error bars from toy measurements
- 1D distributions from profile likelihood

$$\mathcal{L}(\dots, x_{j-1}, x_{j+1}, \dots) \equiv \max_{x_j} \mathcal{L}(x_1, \dots, x_n)$$

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Local view of 2011 data [Klute, Lafaye, TP, Rauch, Zerwas, Dührssen]

- focus on SM solution where possible
- five couplings from data

$g_W \sim 0$ while g_Z okay

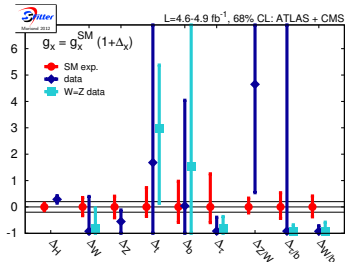
g_b and g_t hurt by secondary solution

g_τ inconclusive in data

g_g and g_γ requiring $t\bar{t}H$ analysis

- poor man's analysis great: $\Delta_j \equiv \Delta_H$

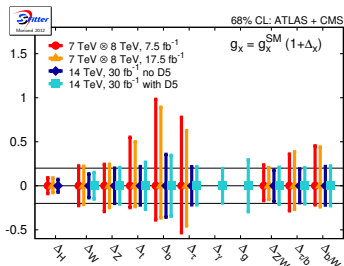
⇒ SFitter at work



Future

2012, 2014, etc

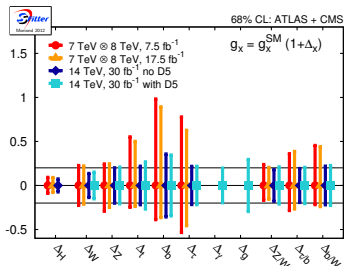
- specially Higgs:
dark side of the Higgs portal?
new states in effective couplings?
 - 2012: meaningful WBF measurements
 g_W and g_τ accessible
 - 2014: $t\bar{t}H$ and $H \rightarrow b\bar{b}$ measurements
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- ⇒ exciting prospects!



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Missing information

- exclusive recoil jet counting [Erik's talk]
- currently unavailable efficiencies [Azatov etal]

N_{ev}/fb	1ℓ	jj	$p_T(\gamma\gamma) < 40 \text{ GeV}$				$p_T(\gamma\gamma) > 40 \text{ GeV}$			
			$R_g^>$ BAR	$R_g^<$ BAR	$R_g^>$ END	$R_g^<$ END	$R_g^>$ BAR	$R_g^<$ BAR	$R_g^>$ END	$R_g^<$ END
GF	0	0.14	3.23	3.40	1.20	1.44	1.55	1.64	0.58	0.69
WBF	0	0.44	0.067	0.071	0.026	0.031	0.17	0.18	0.066	0.079
VH	0.089	0.0035	0.059	0.063	0.028	0.033	0.17	0.18	0.081	0.097
GF/sum	0	0.24	0.96	0.96	0.96	0.96	0.82	0.82	0.80	0.80
WBF/sum	0	0.70	0.02	0.02	0.02	0.02	0.09	0.09	0.09	0.09
VH/sum	1	0.06	0.02	0.02	0.02	0.02	0.09	0.09	0.11	0.11

⇒ a lot published, but pieces missing

Outlook

Data!

Likelihoods

Parameters

Future

Higgs analyses in the present and in the future

- Higgs couplings accessible
- model hypotheses flexible [HDecay]
- technically not trivial [SFitter]
- much more fun than $Z@LEP$
- ‘Why don’t you leave this to experimentalists?’

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- a case for a linear collider!?

Some of this work was funded by the BMBF Theorie-Verbund which is ideal for LHC phenomenology



Bundesministerium
für Bildung
und Forschung

SFitter: Error analysis

Data!

Likelihoods

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Future

Sources of uncertainty

- statistical error: Poisson
systematic error: Gaussian, if measured
theory error: not Gaussian
- simple argument
LHC rate 10% off: no problem
LHC rate 30% off: no problem
LHC rate 300% off: Standard Model wrong
- theory likelihood flat centrally and zero far away
- profile likelihood construction: RFit [CKMFitter]

$$-2 \log \mathcal{L} = \chi^2 = \vec{\chi}_d^T \mathbf{C}^{-1} \vec{\chi}_d$$

$$\chi_{d,i} = \begin{cases} 0 & |d_i - \bar{d}_i| \leq \sigma_i^{(\text{theo})} \\ \frac{|d_i - \bar{d}_i| - \sigma_i^{(\text{theo})}}{\sigma_i^{(\text{exp})}} & |d_i - \bar{d}_i| > \sigma_i^{(\text{theo})} \end{cases}$$

$$|d_i - \bar{d}_i| < \sigma_i^{(\text{theo})}$$

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SFitter: Error analysis

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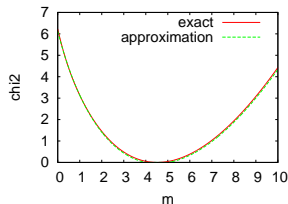
$$|d_i - \bar{d}_i| > \sigma_i^{(\text{theo})}$$

Combination of errors

- Gaussian \otimes Gaussian: half width added in quadrature
- Gaussian/Poisson \otimes flat: RFit scheme
- Gaussian \otimes Poisson: ??
- approximate formula

$$\frac{1}{\log \mathcal{L}_{\text{comb}}} = \frac{1}{\log \mathcal{L}_{\text{Gauss}}} + \frac{1}{\log \mathcal{L}_{\text{Poisson}}}$$

- **modified Minuit gradient fit last step**



SFitter: Markov chains

Data!

Likelihoods

Parameters

Future

Cooling Markov chains [Lafaye, TP, Rauch, Zerwas]

- zoom in on peak structures [inspired by simulated annealing]
- modified condition
Markov chain in 100 partitions, numbered by j

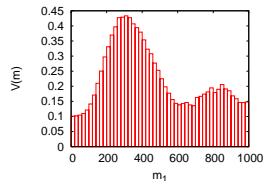
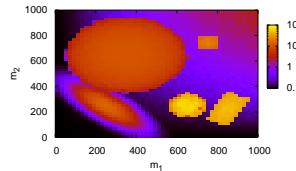
$$\frac{\rho(m')}{\rho(m)} > r^{\frac{100}{c}} \quad \text{with } c \sim 10, \quad r \in [0, 1] \text{ random number}$$

- check for parameter coverage with many Markov chains
- ⇒ **exclusive likelihood map first result**

SFitter: Frequentist vs Bayesian

Getting rid of model parameters

- poorly constrained parameters
- uninteresting parameters
- unphysical parameters [JES part of m_t extraction]
- two ways to marginalize likelihood map
- 1– integrate over probabilities
 - normalization etc mathematically correct
 - integration measure unclear
 - noise accumulation from irrelevant regions
 - classical example: convolution of two Gaussians



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 - 1– integrate over probabilities
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 - classical example: convolution of two Gaussians
 - 2– profile likelihood $\mathcal{L}(\dots, x_{j-1}, x_{j+1}, \dots) \equiv \max_{x_j} \mathcal{L}(x_1, \dots, x_n)$
 - no integration needed
 - no noise accumulation
 - not normalized, no comparison of structures
 - classical example: best-fit point
- one-dimensional parameter distributions second target

