

Towards a systematic UV interpretation of global fits

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Why global fits?

- Need to assess deviations in several observables.
- SMEFT offers a common interpretation to them in a more model-independent way.
- Large number of parameters require broad dataset.
- A truly global fit would be a key piece in the legacy of (HL-)LHC
- However, simple UV models allow us to interpret more easily the results.

Fitmaker

arXiv: 2012.02779

 **SMEFiT**

arXiv: 2105.00006

No interpretation in terms
of UV models yet!

 **SFitter**

arXiv: 2108.01094

The SMEFiT framework

- SMEFT at dimension 6, Warsaw-like basis.
- Datasets: Top quark production, Higgs production and decay, Run II diboson production, LEP WW production, EWPO (approx.).
- State-of-the-art theoretical predictions:
 - SM at NNLO QCD with NLO EW where available.
 - SMEFT predictions with NLO QCD corrections (based on SMEFTatNLO), with interference and quadratic terms.
- Two complementary fitting strategies:
 - MCFit: MonteCarlo replica method, inspired by NNPDF analysis.
 - Nested Sampling: reconstructs the posterior by means of Bayesian inference.

UV assumptions

Any UV model boils down to a restriction of the EFT space.

Applying them to a general fit will fail in general, they must be considered from the beginning.

A simple example

Let's add to the SM another scalar: $\varphi \sim (1, 2)_{1/2} \quad \langle \varphi \rangle = 0$

General UV Lagrangian:

$$\mathcal{L}_{UV} = \mathcal{L}_{SM} + |D_\mu \varphi|^2 - m_\varphi^2 \varphi^\dagger \varphi - \left((y_\varphi^e)_{ij} \varphi^\dagger \bar{e}_R^i \ell_L^j + (y_\varphi^d)_{ij} \varphi^\dagger \bar{d}_R^i q_L^j + (y_\varphi^u)_{ij} \varphi^\dagger i\sigma_2 \bar{q}_L^{T,j} u_R^j + \lambda_\varphi \varphi^\dagger H |H|^2 + h.c. \right)$$

At tree-level, use Granada dictionary! (arXiv: 1711.10391)

We can perform the fit with any one-particle SM extension in there.

Tree-level matching

Flavour symmetry:

$$U(2)_q \times U(2)_u \times U(3)_d \times (U(1)_\ell \times U(1)_e)^3 + \delta y_{b,c,\tau}$$

Enforced at the level of WCs, work out its meaning in terms of UV couplings.

For the example model we're using:

$$\frac{c_{Qt}^{(1)}}{\Lambda^2} = -\frac{(y_{\varphi,33}^u)^2}{6 m_\varphi^2}, \quad \frac{c_{Qt}^{(8)}}{\Lambda^2} = -\frac{(y_{\varphi,33}^u)^2}{m_\varphi^2}, \quad \frac{c_{tH}}{\Lambda^2} = -\frac{y_{\varphi,33}^u \lambda_\varphi}{m_\varphi^2}$$

Why aren't more couplings allowed?

$$0 = (c_{quqd}^1)_{3333} \sim y_{\varphi,33}^u y_{\varphi,33}^d \quad \rightarrow \quad y_{\varphi,33}^d = y_{\varphi,33}^e = 0$$
$$0 = (c_{lequ}^1)_{3333} \sim y_{\varphi,33}^u y_{\varphi,33}^e$$

Simple constraints on WCs

$$\frac{c_{Qt}^{(1)}}{\Lambda^2} = -\frac{(y_{\varphi,33}^u)^2}{6 m_{\varphi}^2},$$

$$\frac{c_{Qt}^{(8)}}{\Lambda^2} = -\frac{(y_{\varphi,33}^u)^2}{m_{\varphi}^2},$$

$$\frac{c_{tH}}{\Lambda^2} = -\frac{y_{\varphi,33}^u \lambda_{\varphi}}{m_{\varphi}^2}$$

Linear relations

$$\frac{c_{Qt}^{(8)}}{\Lambda^2} = 6 \frac{c_{Qt}^{(1)}}{\Lambda^2}$$

Sign-definiteness

$$\frac{c_{Qt}^{(8)}}{\Lambda^2} = -\frac{(y_{\varphi,33}^u)^2}{m_{\varphi}^2} \leq 0$$

The fit is performed assuming all these constraints

Automatized computation for all models

Not so simple constraints on WCs

Relaxed flavour assumptions for the same model give:

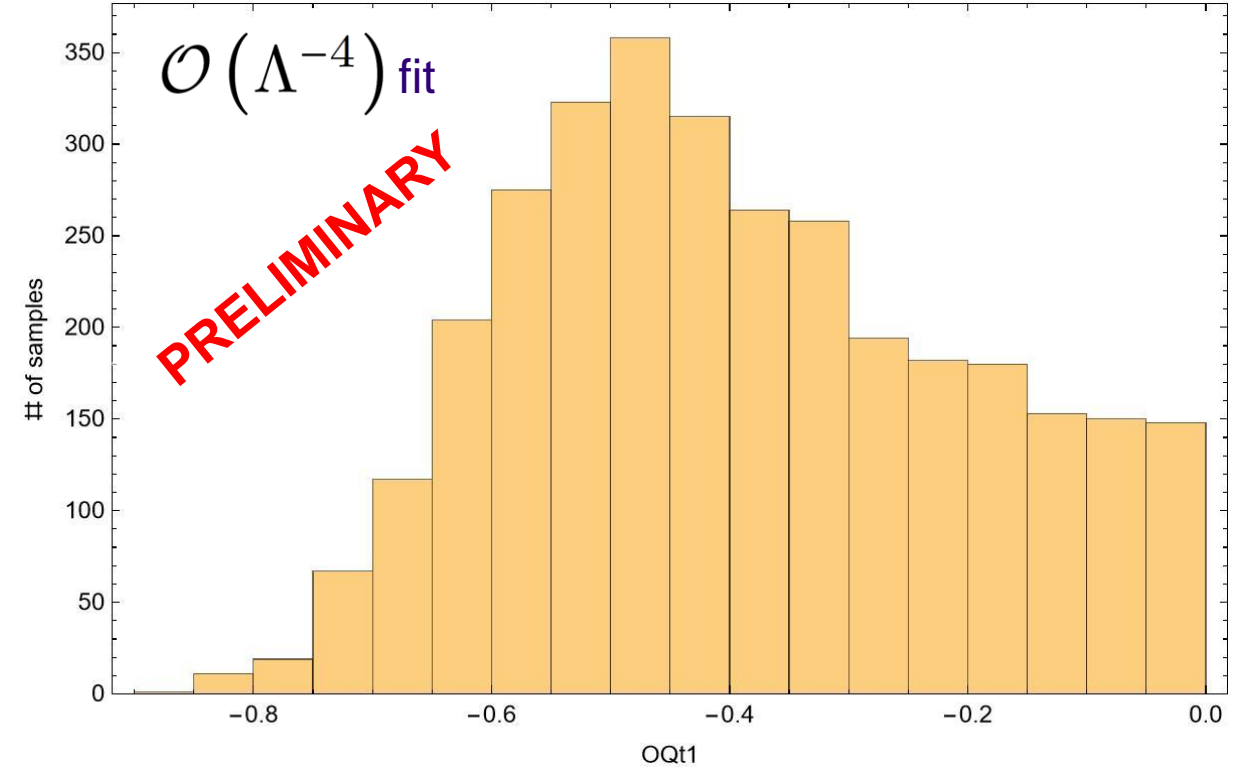
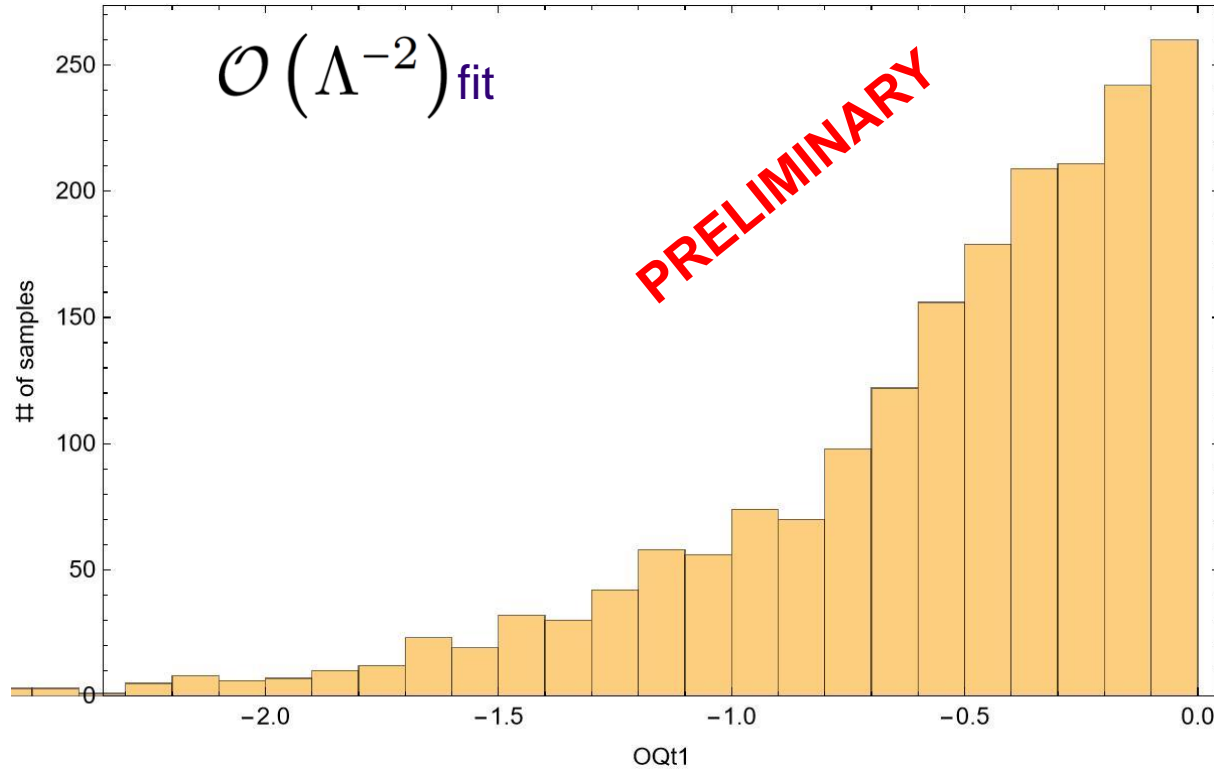
$$c_{qd}^{(1)} = -\frac{\left(\left(y_{\varphi}^d\right)_{33}\right)^2}{6 m_{\varphi}^2} \quad c_{Qt}^{(1)} = -\frac{\left(y_{\varphi,33}^u\right)^2}{6 m_{\varphi}^2} \quad c_{bH} = \frac{\lambda_{\varphi} \left(y_{\varphi}^d\right)_{33}}{m_{\varphi}^2} \quad c_{tH} = -\frac{\lambda_{\varphi} y_{\varphi,33}^u}{m_{\varphi}^2}$$

$$\frac{c_{Qt}^{(1)}}{c_{qd}^{(1)}} = \left(\frac{c_{t\varphi}}{c_{b\varphi}}\right)^2$$

Relations like this are common when using 1-loop matching results.

Their computation is also automatized.

Sign-definite posteriors.



How to correctly compute the bounds with this distribution?

Credible Intervals for bounded distributions

$$\alpha\% \text{ C.I. } \left\{ \begin{array}{l} \text{E.T.I.: } \left[\frac{100-\alpha}{2} \text{th percentile}, \frac{100+\alpha}{2} \text{th percentile} \right] \\ \text{H.D.I.: } \int_{x:p(x)>W} p(x) dx = \frac{\alpha}{100} \end{array} \right.$$

E.T.I.: Equal-tailed interval

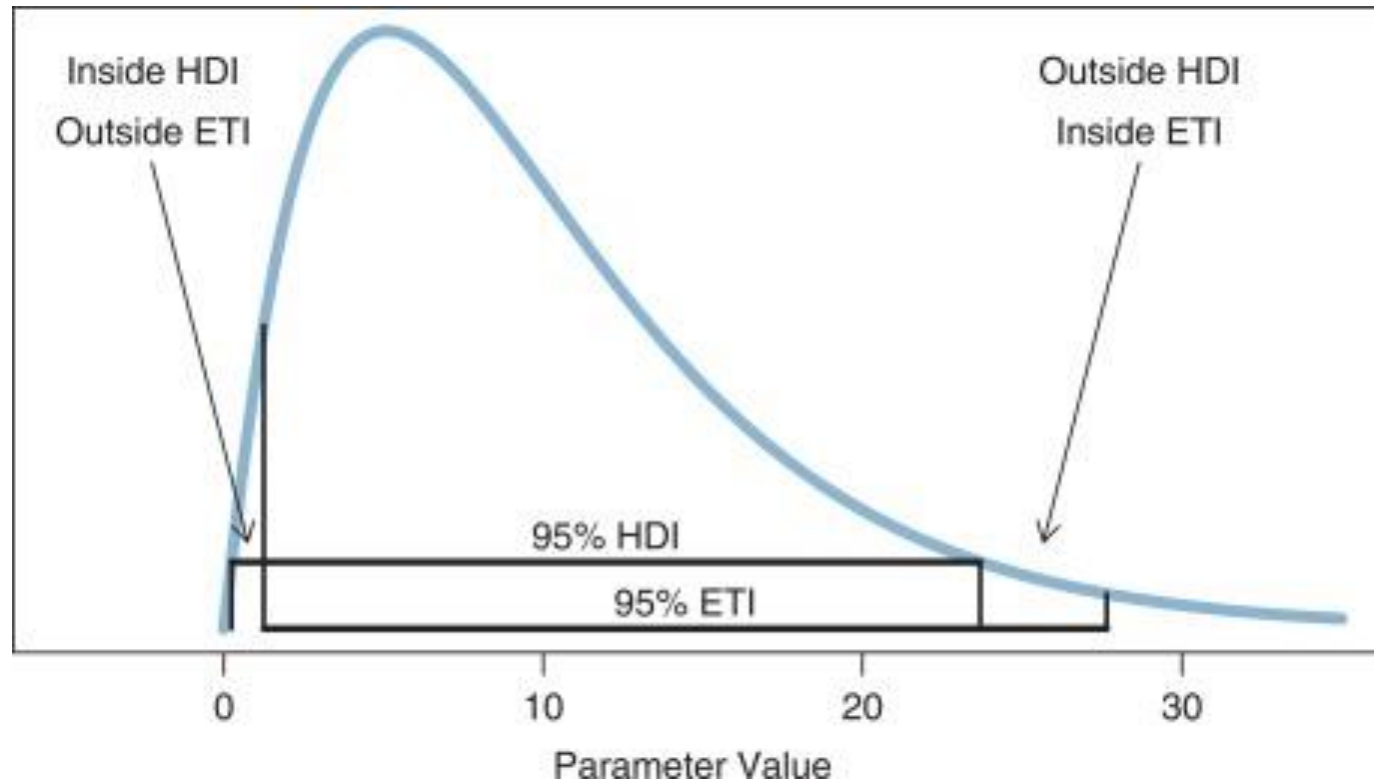
H.D.I.: Highest-density interval

Adapting the ETIs to bounded distributions

WC sign property	95% C.I. definition
Unrestricted	(2.5th percentile, 97.5th percentile)
Positive defined	[0, 95th percentile)
Negative defined	(5th percentile, 0]

Reference: J. K. Kruschke, *Doing Bayesian Data Analysis*, Second Ed., Academic Press (2015) Ch. 12, Pages 335-358.

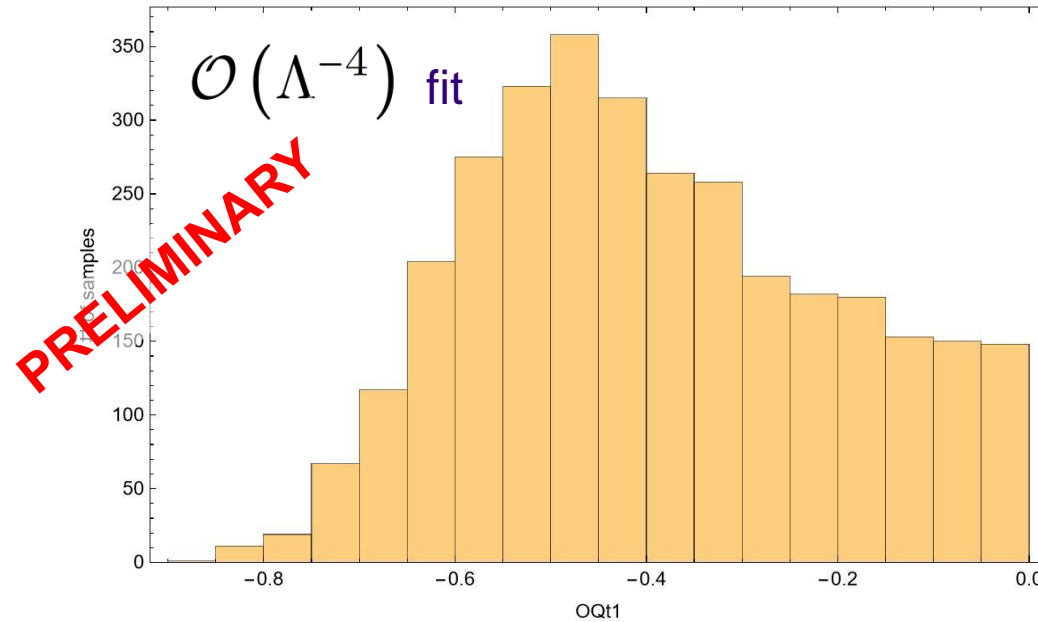
Credible Intervals for bounded distributions



$$\alpha\% \text{ C.I. } \left\{ \begin{array}{l} \text{E.T.I.: } \left[\frac{100-\alpha}{2} \text{th percentile, } \frac{100+\alpha}{2} \text{th percentile} \right] \\ \text{H.D.I.: } \int_{x:p(x)>W} p(x) dx = \frac{\alpha}{100} \end{array} \right.$$

Image from: J. K. Kruschke, *Doing Bayesian Data Analysis*, Second Ed., Academic Press (2015) Ch. 12, Pages 335-358.

Credible Intervals for bounded distributions



Comparing results with ETIs and HDIs

PRELIMINARY


	68% C.I.	95% C.I.	68% HDI	95% HDI
$c_{Qt}^{(1)}$	$(-0.499, 0]$	$(-0.672, 0]$	$(-0.633, -0.238)$	$(-0.690, -0.017)$
$c_{Qt}^{(8)}$	$(-2.992, 0]$	$(-4.028, 0]$	$(-3.799, -1.43)$	$(-4.140, -0.100)$
$c_{t\varphi}$	$(-0.908, 0.063)$	$(-1.285, 0.352)$	$(-0.902, 0.042)$	$(-1.314, 0.352)$


What UV information can we extract?

$$\frac{c_{Qt}^{(1)}}{\Lambda^2} = -\frac{(y_{\varphi,33}^u)^2}{6 m_\varphi^2},$$

$$\frac{c_{Qt}^{(8)}}{\Lambda^2} = -\frac{(y_{\varphi,33}^u)^2}{m_\varphi^2},$$

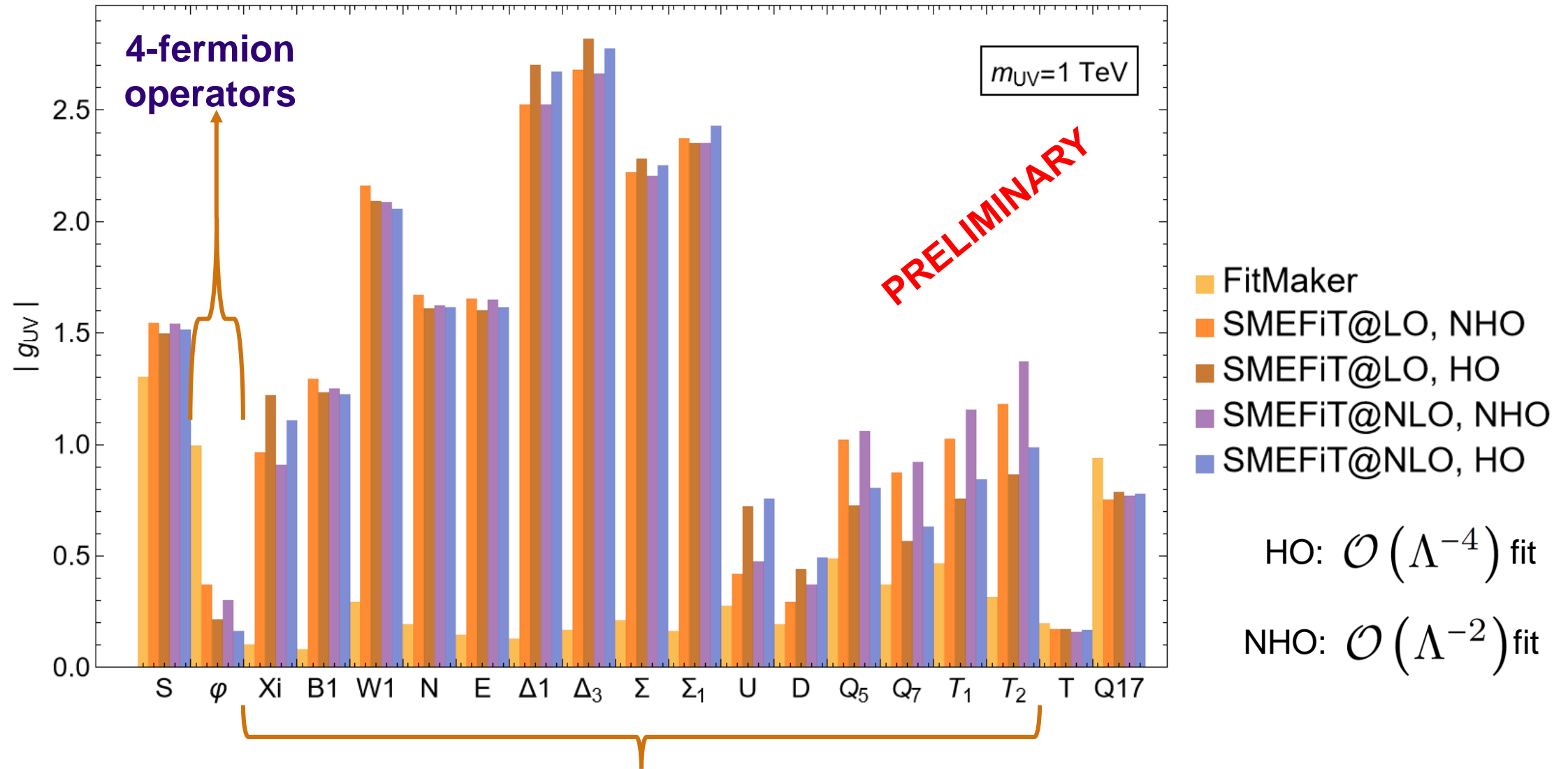
$$\frac{c_{tH}}{\Lambda^2} = -\frac{y_{\varphi,33}^u \lambda_\varphi}{m_\varphi^2}$$


$$|y_{\varphi,33}^u|, \quad \lambda_\varphi \frac{y_{\varphi,33}^u}{|y_{\varphi,33}^u|}$$


$$\text{sgn}(\lambda_\varphi y_{\varphi,33}^u) \times |\lambda_\varphi|$$

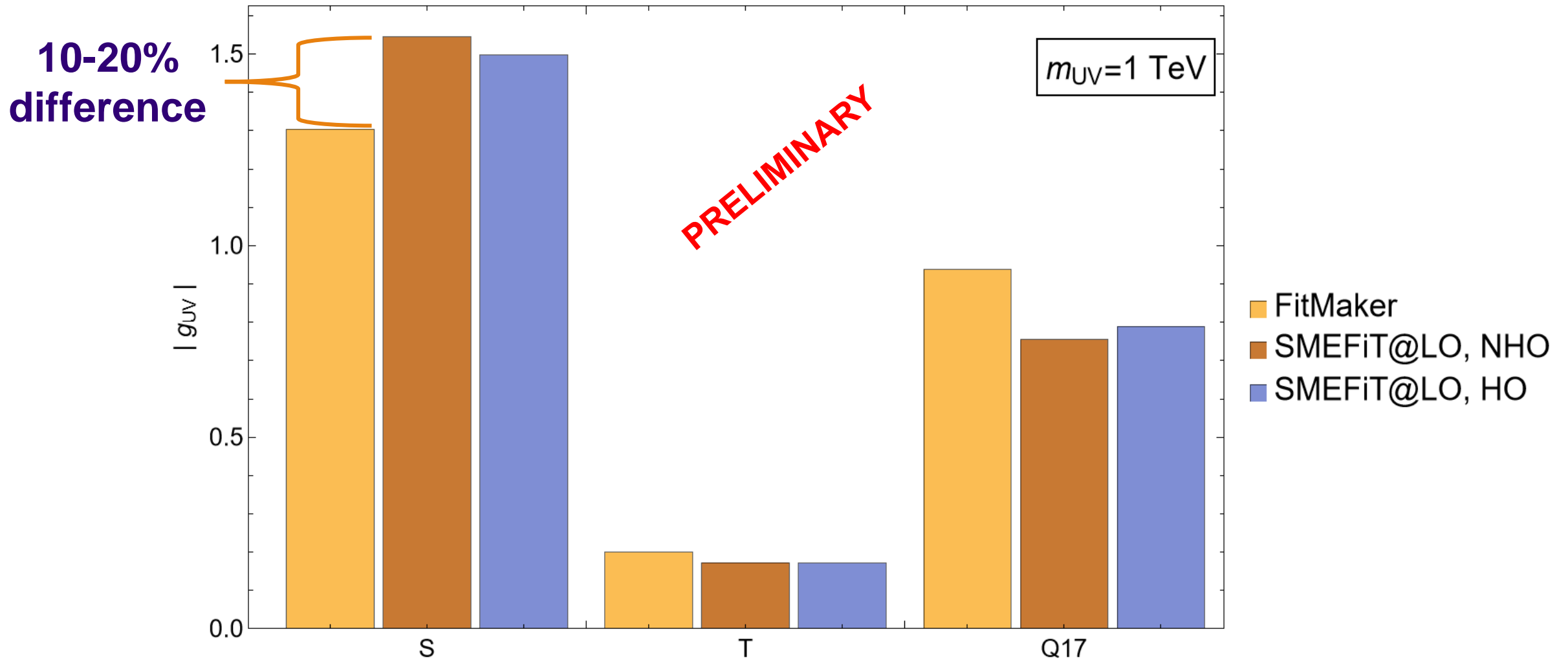
Automatized computation.

Comparison with FitMaker



Constrained by EWPOs

Comparison with FitMaker



Models for which the comparison is fair.

1-loop matching, our next milestone

- Use of Matchmaker EFT to compute the 1-loop matching results.
- We have the matching results for a handful of models.
- A big difficulty could be imposing the relations among WCs...

$$\left(y_{\varphi,33}^u\right)^2 = \frac{\left(\lambda_{\varphi} y_{\varphi,33}^u\right)^2}{\left(\lambda_{\varphi}\right)^2} \longrightarrow 0 = \left(-\frac{g_2^3}{320 g_3^2} \frac{c_{tq}^{(8)}}{c_{WWW}}\right) \left(\frac{27}{64 g_3^2} c_{tq}^{(8)} + \frac{720}{g_2^3} \frac{(18653761 + 93268136\pi^2)}{11658517} c_{WWW}\right)^2 + \frac{\left(c_{t\varphi} - \frac{725090343831}{1865864036231} \frac{1}{g_3^2 \sqrt{2}} c_{tq}^{(8)} + \frac{4319\sqrt{2}}{12311} \frac{g_1^4 + 6g_2^4}{g_2^3} c_{WWW} + \frac{17276\sqrt{2}}{36933} c_{\varphi d}\right)^2}{192 \left(\frac{g_1^4}{138240} + \frac{g_2^4}{46080}\right) + \frac{g_2^3}{540} \frac{c_{\varphi d}}{c_{WWW}}},$$

- But we're working on an alternative route!

Conclusion and outlook

- **Volume of data calls for general and automatized analysis/fitting frameworks.**
- **The inclusion of UV models helps to understand the meaning of the fits.**
- **SMEFiT is on the way towards a very general and automatized framework to do this.**
- **The implementation with tree-level matching is mostly ready.**
- **The 1-loop matching case will be ready soon.**
- **The inclusion of EWPOs in SMEFiT is a pressing issue.**

Thank you for your attention

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