

Beyond the Standard Model : phenomenology and LHC constraints of exotic dark matter

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LPSC Grenoble & LAPTh Annecy

Mid-thesis presentation, March 25, 2021

Supervised by Sabine Kraml and Genevieve Belanger

- 1 From dark matter to LHC
- 2 Setting constraints with SModelS and MadAnalysis5
- 3 Probing exotic dark matter (work in progress)

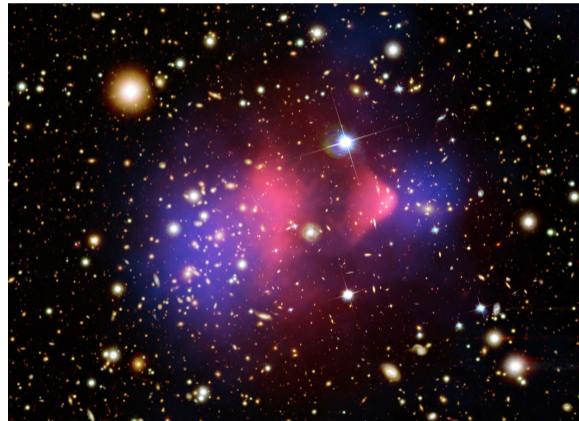
G.A., Araz, Fuks, Kraml, Waltenberger, Les Houches 2019 proceedings, contribution 15 [\[arXiv\]](#)

G.A., Kraml, Waltenberger, CPC 264 (2021) 107909 [\[arXiv\]](#)

G.A., Heisig, Khosa, Kraml, Kulkarni, Lessa, Neuhuber,

Reyes-González, Waltenberger, Wongel, TOOLS 2020 proceedings [\[arXiv\]](#)

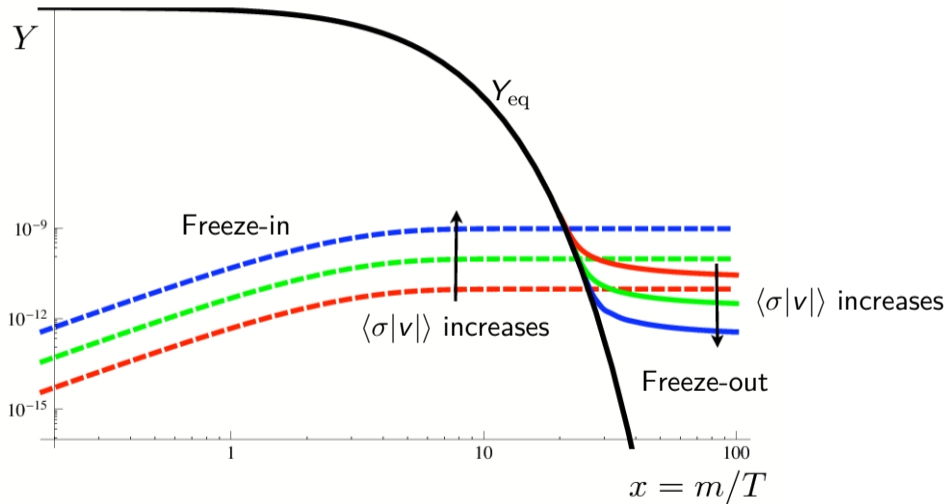
- Many evidences hinting for a missing mass
 - Galaxy rotation curves
 - Galaxy clusters
 - Gravitational lensing
 - And others
- Must be massive, weakly interacting, dark

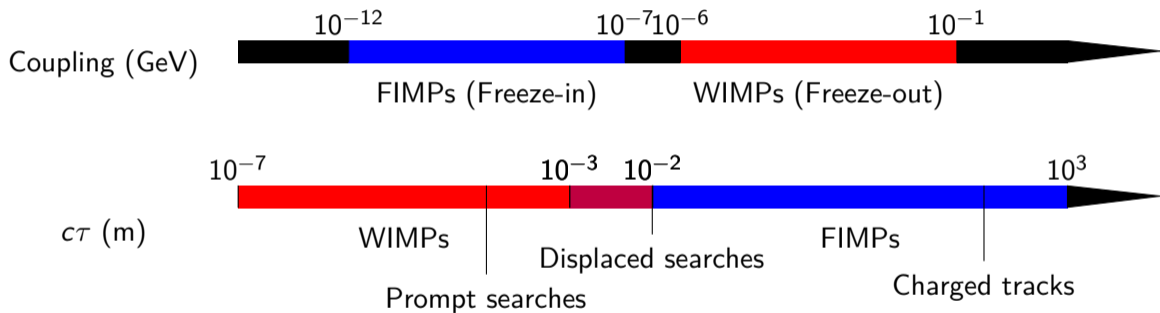


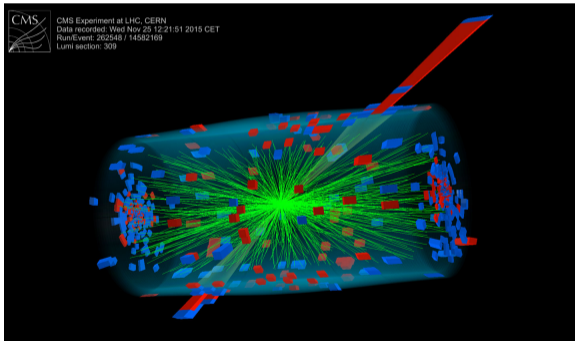
The DM theory landscape



Nature 562, 5156 (2018)

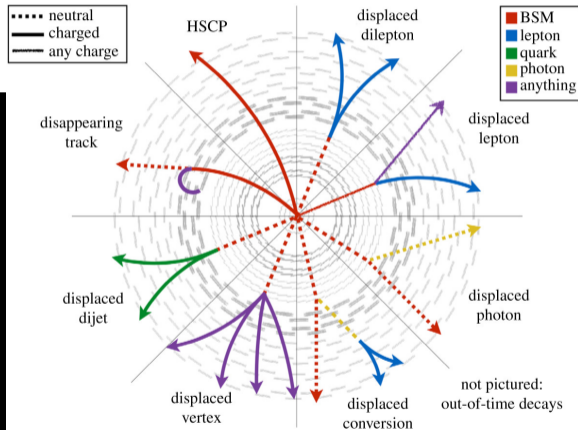






CMS

CMS Experiment at LHC, CERN
Data recorded: Wed Nov 25 12:21:51 2015 CET
Run/Event: 282548 / 14582169
Lumi section: 309



- Monte Carlo event generation + detector simulation (MadGraph + MadAnalysis)
 - simulate the whole collision, detector effects and selection cutflow
 - captures differences in kinematical distributions
 - more generally applicable
 - very CPU-time consuming
- Using simplified model results (SModelS)
 - uses the efficiencies provided by experimentalists
 - assumes the cut acceptances are approx. the same as in the simplified models
 - much faster
 - suitable for large scans

G.A., Araz, Fuks, Kraml, Waltenberger, Les Houches 2019 proceedings, contribution 15 [\[arXiv\]](#)

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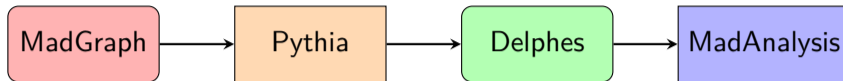
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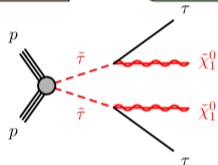
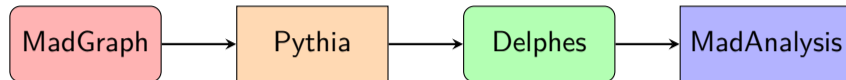
G.A., Kraml, Waltenberger, CPC 264 (2021) 107909 [\[arXiv\]](#)

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ATLAS-SUSY-2018-04

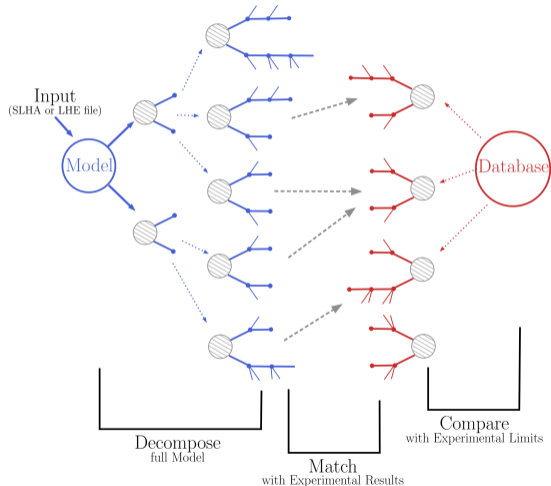


SR – lowMass	SR – highMass
2 tight τ (OS)	2 medium τ (OS)
–	≥ 1 tight τ
$\Delta R(\tau_1, \tau_2) < 3.2$	$E_T^{\text{miss}} > 150 \text{ GeV}$
$75 < E_T^{\text{miss}} < 150 \text{ GeV}$	$\Delta R(\tau_1, \tau_2) < 3.2$
$m_{T2} > 70 \text{ GeV}$	$m_{T2} > 70 \text{ GeV}$

ATLAS-SUSY-2018-04

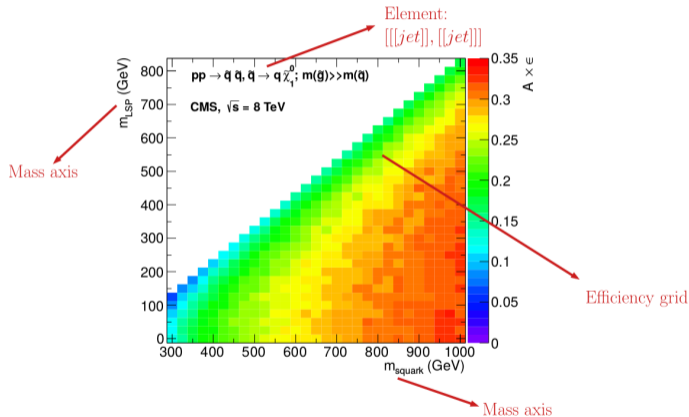
- Based on a general procedure to decompose BSM collider signatures featuring a \mathbb{Z}_2 -like symmetry into simplified-model topologies
- Large database of simplified-model results (currently ~ 100 ATLAS and CMS searches)
- New generic treatment of width-dependent results with a variety of LLP analyses

See online SModelS documentation at
smodels.readthedocs.io



- Upper limit (UL)
 - Constrains $\sigma \times \prod_j BR_j$
 - Only binary decision

- Efficiency map (EM)
 - Allows to sum contributions from several topologies
 - Can compute a likelihood (confidence level)



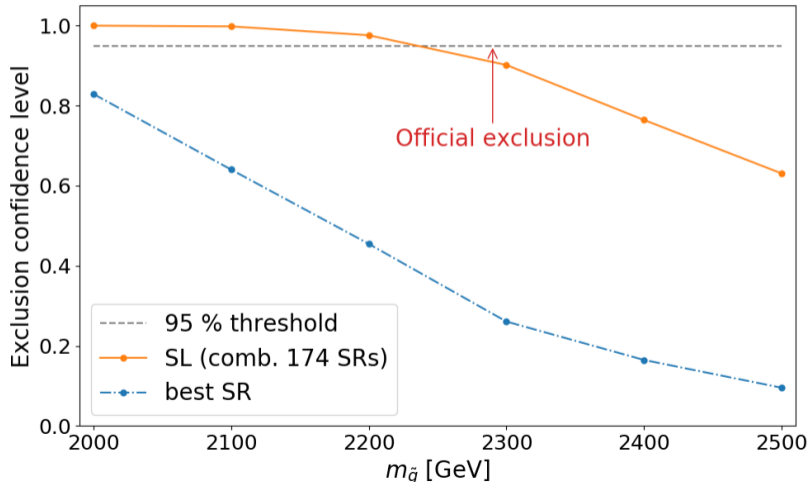
- For a proper statistical evaluation
 - need to compute a likelihood
 - without specific likelihood information : can only compute for each signal region (SR)

$$\mathcal{L}(\mu, \theta | D) = \frac{(\mu s + b + \theta)^{n_{obs}} e^{-(\mu s + b + \theta)}}{n_{obs}!} \exp\left(-\frac{\theta^2}{2\delta^2}\right)$$

⇒ use the “most sensitive” SR

- **Simplified likelihoods** : global background uncertainty [[CMS-NOTE-2017-001](#)]
 - summarized into a covariance matrix
 - one Gaussian error for each SR
- **Full likelihoods** : full statistical descriptions of analyses [[ATL-PHYS-PUB-2019-029](#)]
 - encapsulate the detailed information about the analysis (detailed systematic uncertainties)
 - reproduces exactly the experimental analysis and allows for a more precise reinterpretation

CMS-SUS-19-006: $pp \rightarrow \tilde{g}\tilde{g}, \tilde{g} \rightarrow b\bar{b}\tilde{\chi}_1^0$ ($m_{\tilde{\chi}_1^0} = 200$ GeV)



Les Houches 2019
proceedings, contribution
15 [\[arXiv\]](#)

- CMS-SUS-16-039
- CMS-SUS-17-001
- CMS-SUS-19-006
- CMS-SUS-16-048

- Serialization of HistFactory workspaces under the JSON format
- Provides observed signals, expected backgrounds and systematic uncertainties as in the experimental analysis
- Can be used in RooFit or [pyhf](#) (a pure-python implementation of HistFactory)

[ATL-PHYS-PUB-2019-029](#)

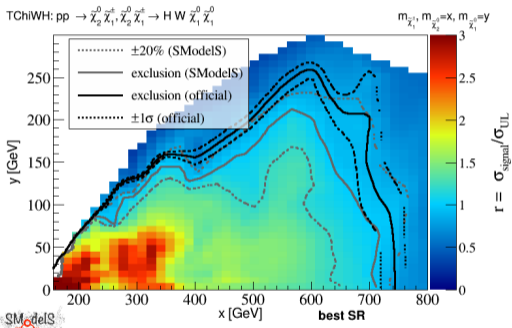
```
{
  "channels": [
    {
      {
        {
          "name": "SR1cut_cuts"
          "samples": [
            {
              "name": "SR2cut_cuts"
              "samples": [
                {
                  "data": [
                    2.570836067199707
                  ],
                  "modifiers": [
                    {
                      "name": "Boson_Staus"
                    }
                  ]
                },
                {
                  "data": [
                    0.044047050178050995
                  ],
                  "modifiers": [
                    {
                      "name": "Z_Staus"
                    }
                  ]
                }
              ]
            }
          ]
        }
      }
    }
  ]
}
```

Signal regions

Background contributions

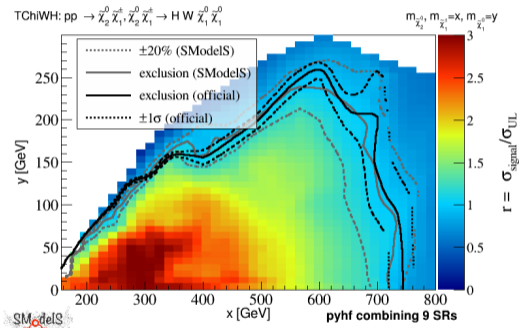
CPC 264 (2021) 107909 [\[arXiv\]](#) and TOOLS2020 proceedings [\[arXiv\]](#)

ATLAS-SUSY-2019-08 (efficiencyMap)



Standard “best SR” procedure

ATLAS-SUSY-2019-08 (efficiencyMap)



pyhf likelihoods

⇒ ATLAS-SUSY-2018-04, ATLAS-SUSY-2018-31 and more to come

USAGE IN PHENO RECASTING TOOLS

Usage in MadAnalysis 5

G. Alguero, J. Araz, B. Fuks, SK, W. Waltenberger
Conf. 15, LH 2019 BSM WG report, arXiv:2002.12220

MA5-pyhf interface established within the LH PhysTeV 2019 workshop

- the relevant JSON files must be located in the same analysis folder as the recast code (done automatically at the time of the PAD installation).
- The `analysis.info` file must include new `<pyhf>` elements specifying the names of the JSON files together with the corresponding channels (ensembles of SRs) and the regions they include, as defined in the JSON files.

```
<pyhf id="RegionA">
  <name>atlas_susy_2018_031_SRA.json</name>
  <regions>
    <channel name="SR_eff">SRA_L_SRA_M_SRA_H</channel>
    <channel name="Vrtt_eff"> </channel>
    <channel name="CRtt_eff"> </channel>
  </regions>
</pyhf>
```



SModelS-pyhf interface

Gaël Alguero, SK, Wolfgang Waltenberger,
arXiv:2008.01809

- Available from SModelS v1.2.4 onward (released Sep. 3rd, 2020)
- The interfacing of pyhf to SModelS consists of two parts:
 - addition of an independent module `tools/pyhfInterface.py`
 - changes brought to `experiment/datasetObj.py`
- Can be turned on/off by setting

```
combineSR = True/False
```

in the `parameters.ini` file ↴

PyhfData class:
Storing and handling of the information related to the JSON files and input signal predictions.
Collects information in the workspaces such as the number of SRs, and the paths to the SR's `workspace` where the BSM predictions are to be written.
The VFs and CFs are assumed not to contribute and removed from the workspaces.

PyhfUpperLimitComputer class:
For inferring the upper limits given the PyhfData information

*) The same flag also turns on the SR combination in the simplified likelihood approach for CMS efficiency map results, for which a covariance matrix is available.

EP-IT Data science seminars

PHYSTAT seminar: Likelihood publishing, RECAST, and simulation-based inference

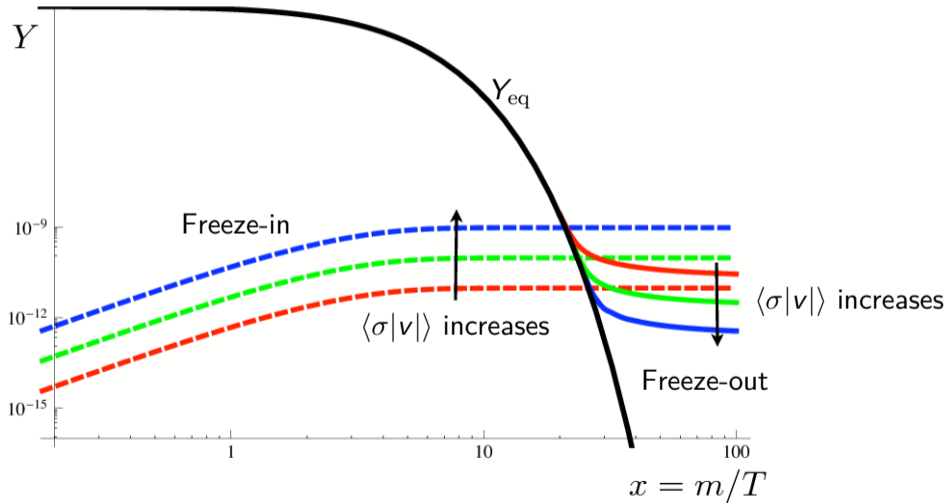
by Kyle Stuart Cranmer (New York University (US))

📅 Wednesday 14 Oct 2020, 14:00 → 15:00 Europe/Zurich

📍 CERN

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“In between” freeze-out and freeze-in



- Standard freeze-out equation

$$\dot{n}_\chi + 3Hn_\chi = -\langle\sigma|v|\rangle (n_\chi^2 - n_{\chi,\text{eq}}^2)$$

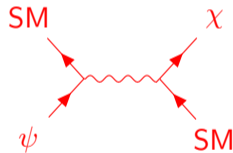
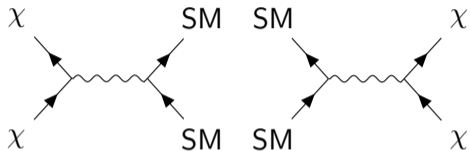
- Freeze-in

- coscattering

$$\dot{n}_\psi + 3Hn_\psi = -\langle\sigma|v|\rangle (n_\psi^2 - n_{\psi,\text{eq}}^2) -$$

$$\Gamma_{\psi\rightarrow\chi} \left(n_\chi - n_\psi \frac{n_\chi^{\text{eq}}}{n_\psi^{\text{eq}}} \right)$$

$$\dot{n}_\chi + 3Hn_\chi = \Gamma_{\psi\rightarrow\chi} \left(n_\chi - n_\psi \frac{n_\chi^{\text{eq}}}{n_\psi^{\text{eq}}} \right)$$



χ : dark matter

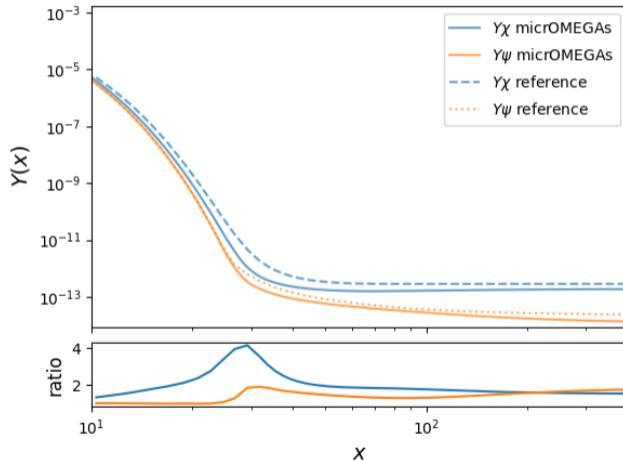
ψ : heavier odd particle

R.T.D'agnolo et al., Phys. Rev. Lett. 119, 061102 (2017) [\[arXiv\]](#)

M.Garny et al., Phys. Rev. D 96, 103521 (2017) [\[arXiv\]](#)

- Testing coscattering with a Singlet-Triplet model
 - Standard Model $+\chi^0 + \{\psi^0, \psi^\pm\}$
- with a compressed spectrum
 - small mass splitting
 - small couplings

$$M = 500 \text{ GeV}, m = 490 \text{ GeV}, \theta = 1.22 \times 10^{-6}$$



F.Brümmer, JHEP 2001 (2020) 113

[\[arXiv\]](#)

Ernest Ma, Phys.Rev.D73:077301,2006 [\[arXiv\]](#)

- Standard Model extended with
 - an inert Higgs doublet
 - and right-handed neutrinos

$$\begin{pmatrix} H_1^\pm \\ H_1^0 \end{pmatrix}, \begin{pmatrix} H_2^\pm \\ H_2^0 \end{pmatrix}, \begin{pmatrix} N_1 \\ N_2 \\ N_3 \end{pmatrix}$$

- odd under a \mathbb{Z}_2 -symmetry

Field	Generations	$SU(3)_c$	$SU(2)_L$	$U(1)_Y$	\mathbb{Z}_2
ℓ_L	3	1	2	$-1/2$	+
e_R	3	1	1	-1	+
H_1	1	1	2	$1/2$	+
H_2	1	1	2	$1/2$	-
N	3	1	1	0	-

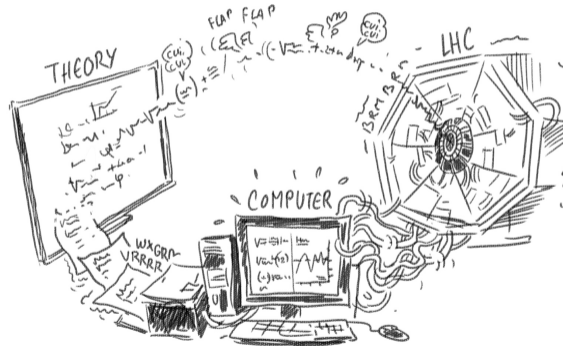
$$H_1 \xrightarrow{\mathbb{Z}_2} H_1$$

$$H_2 \xrightarrow{\mathbb{Z}_2} -H_2$$

- radiative neutrino masses
- provides different DM candidates
 - different production mechanisms (freeze-in, freeze-out, ...)
 - probe with the appropriate LHC signatures (prompt, long-lived, ...)

- simplified likelihoods implemented in MadAnalysis5
 - ability to use CMS covariance matrices
 - database extended with all usable covariance matrices
- SModelS/pyhf interface from SModelS v1.2.4 onwards
 - ability to use full likelihoods from ATLAS
 - added EMs + JSON files for three 139 fb^{-1} analyses
- more accurate reinterpretations, more reliable statistical evaluation e.g. for fits
- work in progress
 - cospattering in micrOMEGAs
 - scotogenic model

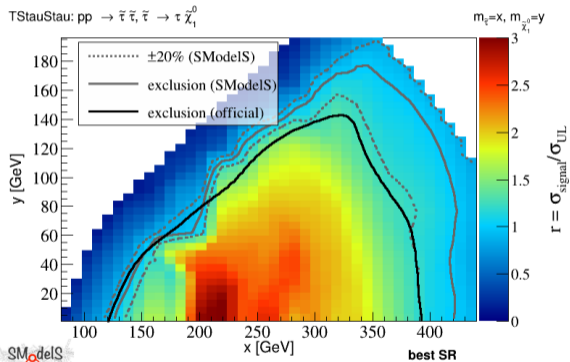
Thanks for your attention !



©Lison Bernet

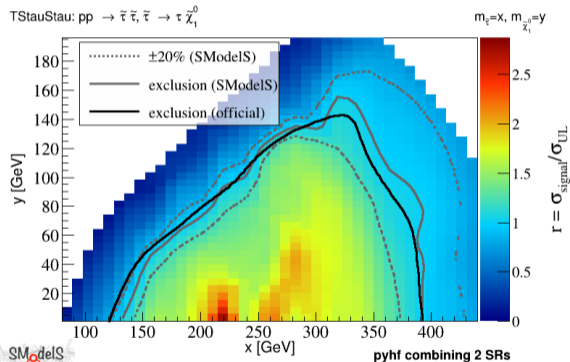
arXiv:2009.01809 (pyhf interface)

ATLAS-SUSY-2018-04 (efficiencyMap)



Standard "best SR" procedure

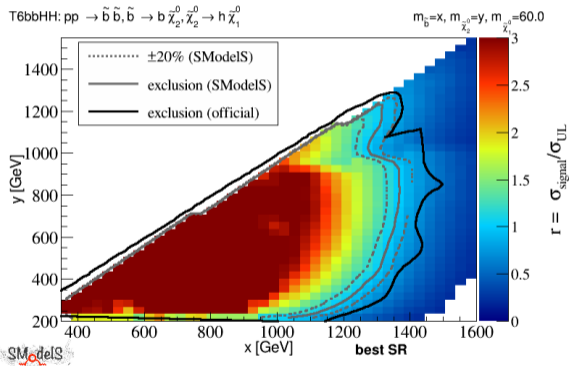
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pyhf likelihoods

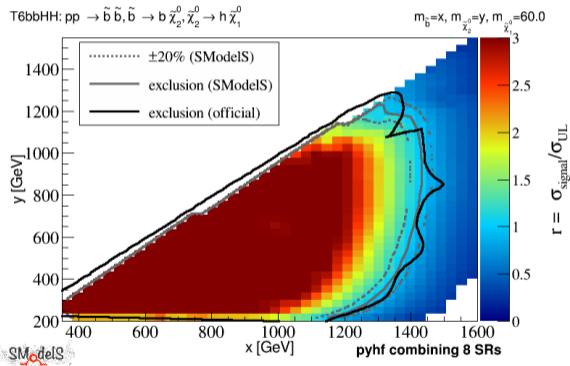
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ATLAS-SUSY-2018-31 (efficiencyMap)



Standard "best SR" procedure

ATLAS-SUSY-2018-31 (efficiencyMap)



pyhf likelihoods

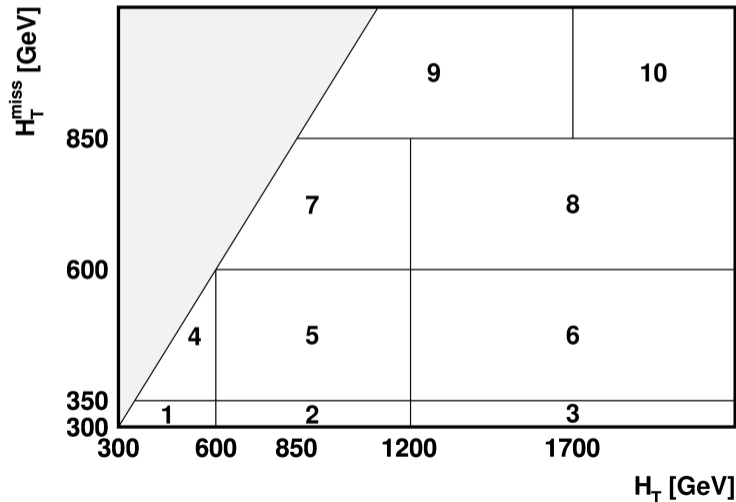
$$p(n, a|\eta, \chi) = \underbrace{\prod_{c \in \text{channels}} \prod_{b \in \text{bins}} \text{Pois}(n_{cb}|\nu_{cb}(\eta, \chi))}_{\text{Measurements}} \underbrace{\prod_{\chi} c_{\chi}(a_{\chi}|\chi)}_{\text{Constraints}} \quad (1)$$

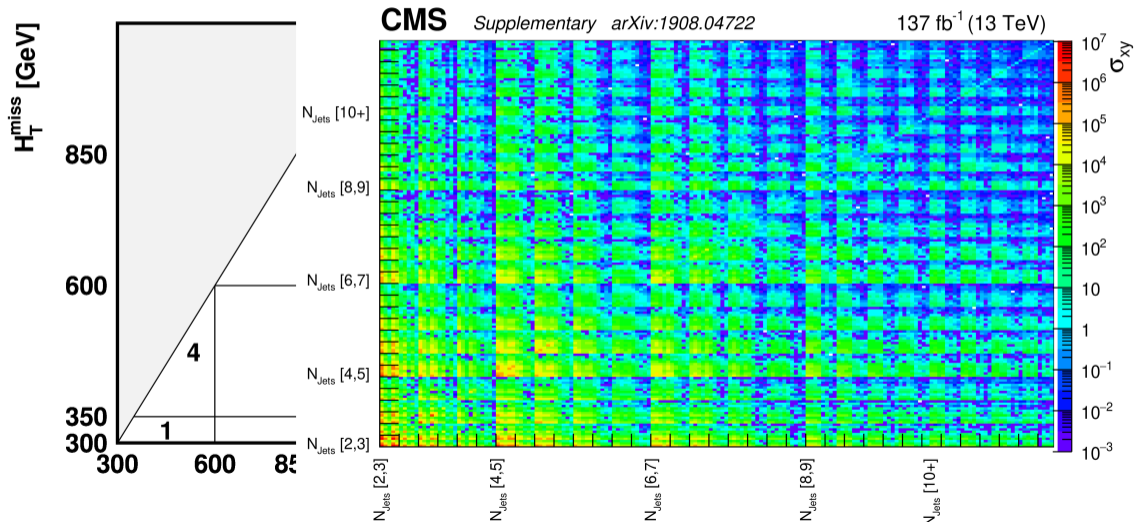
relates the observed events and auxiliary data (n, a) to the free and constrained parameters (η, χ)

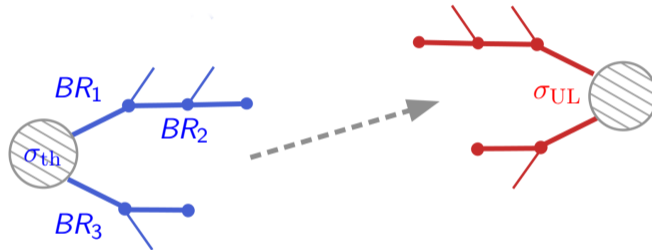
$$\nu_{cb}(\eta, \chi) = \sum_{s \in \text{samples}} \nu_{scb}(\eta, \chi) = \sum_{s \in \text{samples}} \underbrace{\prod_{\kappa} \kappa_{scb}(\eta, \chi)}_{\text{Multiplicative modifiers}} \left(\nu_{scb}^0(\eta, \chi) + \underbrace{\sum_{\Delta} \Delta_{scb}(\eta, \chi)}_{\text{Additive modifiers}} \right) \quad (2)$$

$$\mathcal{L}(\mu, \theta | D) = \prod_{i=1}^N \frac{(\mu s_i + b_i + \theta_i)^{n_{obs}^i} e^{-(\mu s_i + b_i + \theta_i)}}{n_{obs}^i!} \exp\left(-\frac{1}{2} \vec{\theta}^T V^{-1} \vec{\theta}\right)$$

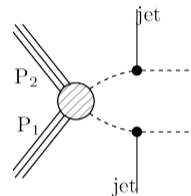
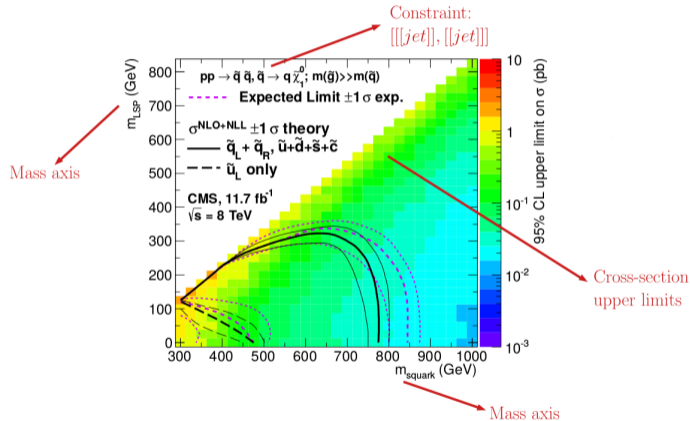
with a product over all N SRs where V is the covariance matrix



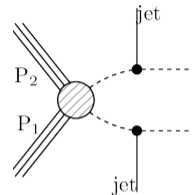
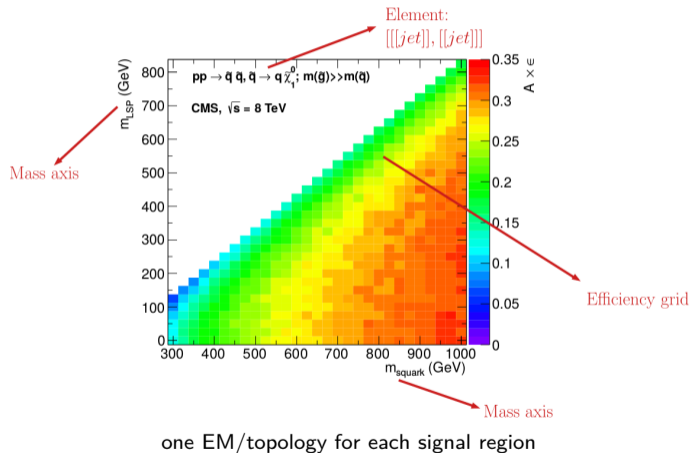




$\sigma_{th} \times BR_1 \times BR_2 \times BR_3$ to compare with σ_{UL}



- Constrains $\sigma \times \prod_j BR_j$ per topology
- Only binary decision : excluded or not



- Constrains $\sum_i A_i \epsilon_i \times \sigma_i \times \prod_j BR_j$ per signal region
- Can sum contributions from several topologies
- Can compute a likelihood (exclusion confidence level)

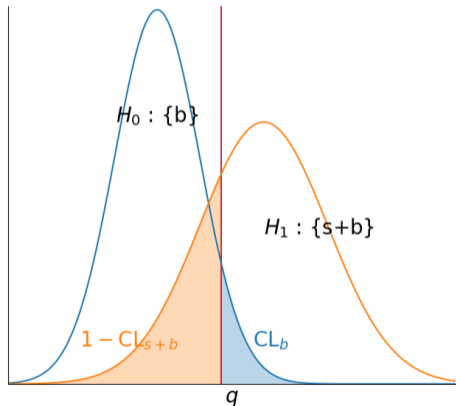
Efficiency map

- Testing hypothesis H_0 : {only background} against H_1 : {background + new signal}
- Definition of a log likelihood ratio

$$q = -2 \log \left(\frac{\mathcal{L}(\text{data}|H_1)}{\mathcal{L}(\text{data}|H_0)} \right)$$

- Computing a robust p -value ratio

$$CL_s = \frac{CL_{s+b}}{CL_b}$$



$$N = A\epsilon\sigma\mathcal{L}$$

- A and ϵ : acceptance and efficiency
- σ and \mathcal{L} : cross section and integrated luminosity