

Prelude

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Let's for a moment entertain the idea of a best case scenario:

we see signs of new physics in several final states at the LHC.

What would we do?

Prelude

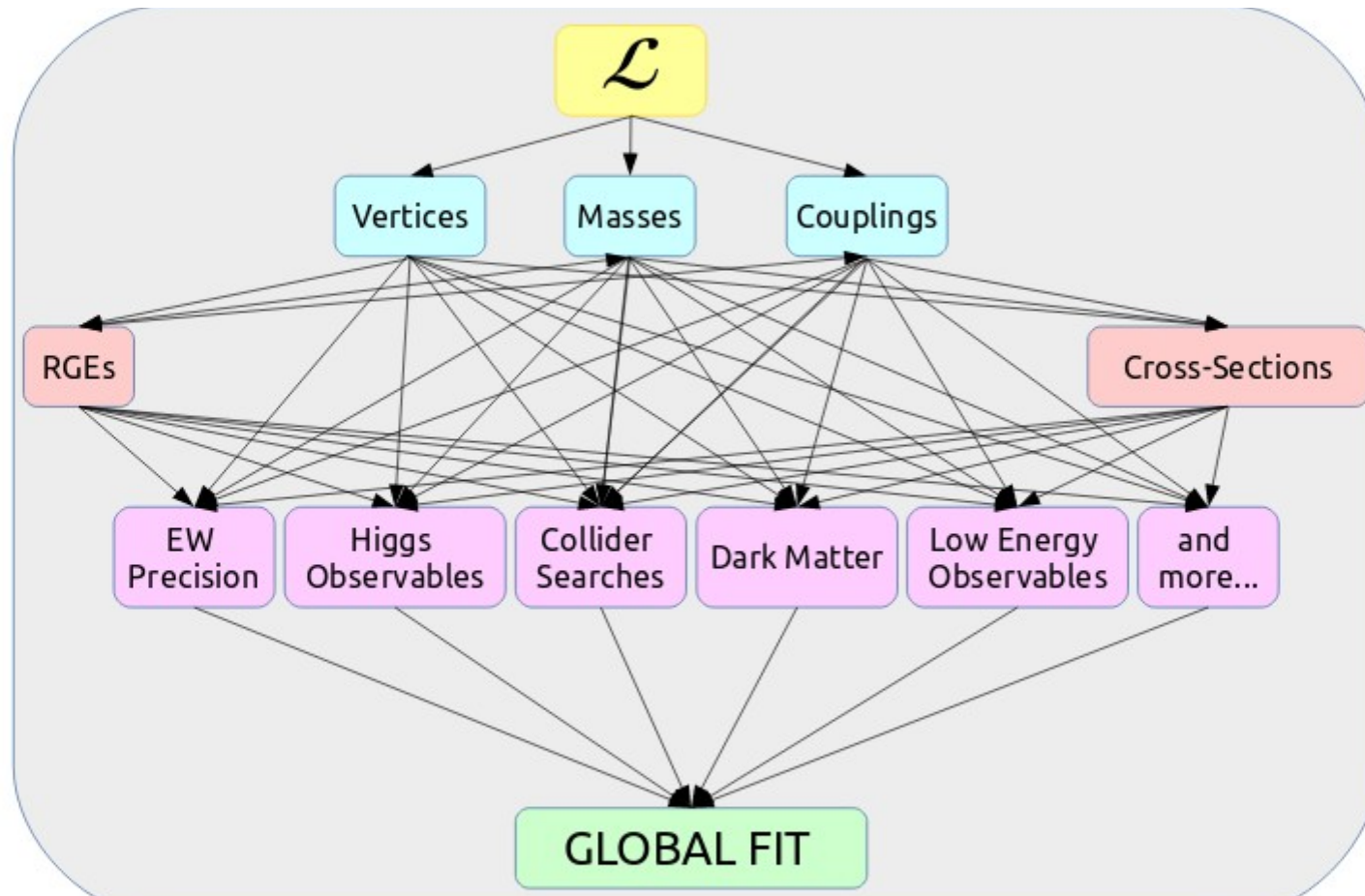
From Jamie's slides [*] (“top-down” approach):

- **We would write our favorite Lagrangian down**
- **We would press “enter”**
- **The computer would fit the model parameters to all relevant observables!**

[*] Would we do the same for positive results?

Prelude

From Jamie's slides:



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I would want to work on a “bottom-up” approach:

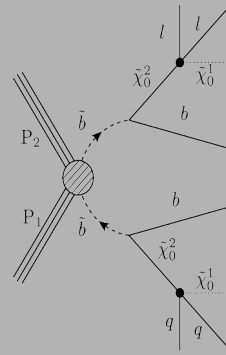
- I would try to go bottom-up, starting from the data.
- I would tackle every excess individually.
- First I would try to find out the set of simplified models that best describes the data using Bayes factors.
- Then I would estimate the model parameters. **The result would be a statistical description of a signal excess in the context of simplified models spectra:**

(From a talk I gave about the description of signal excesses using Bayes factors)

**MOCKUP! Not real data,
not even real simulation.
assuming binned data.**

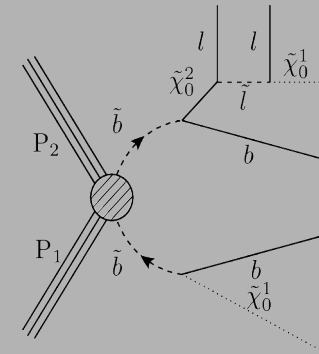
Signal model 1

e.g.

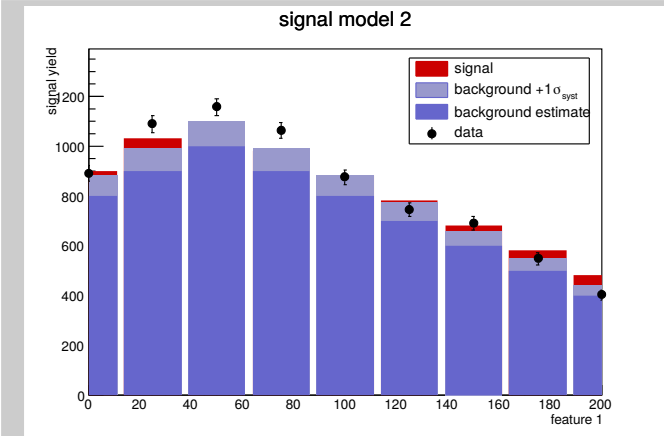
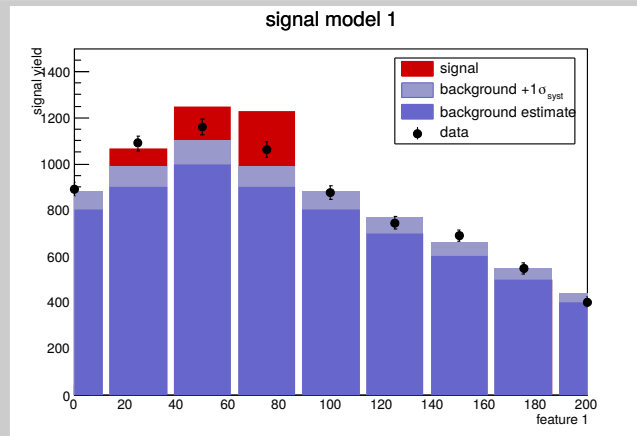


Signal model 2

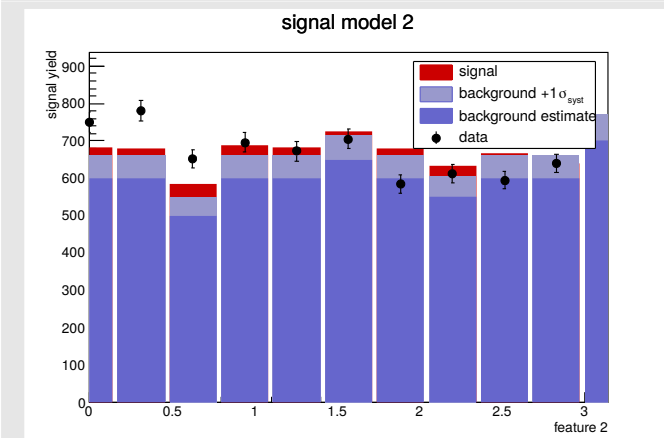
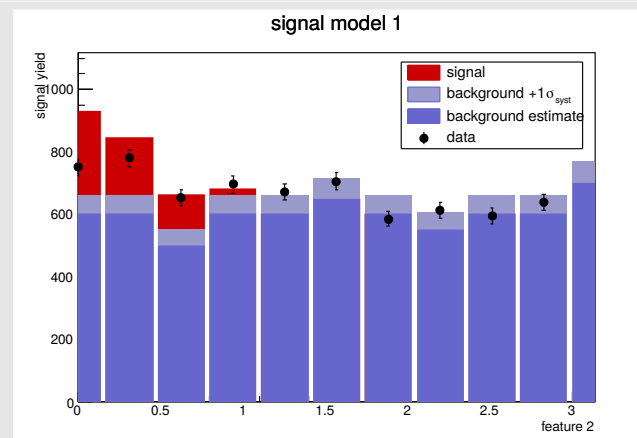
e.g.



Feature 1 (e.g. „mll“)



Feature 2 (e.g. „\phi(l)“)



(From a talk I gave about the description of signal excesses using Bayes factors)

Bayes factors

- Bayesian alternative to frequentist hypothesis testing
- Given a model selection problem in which we have to choose between two models:

$$K = \frac{P(D|M_1)}{P(D|M_2)} = \frac{\int P(\theta_1|M_1)P(D|\theta_1, M_1)d\theta_1}{\int P(\theta_2|M_2)P(D|\theta_2, M_2)d\theta_2}$$

K = Bayes Factor

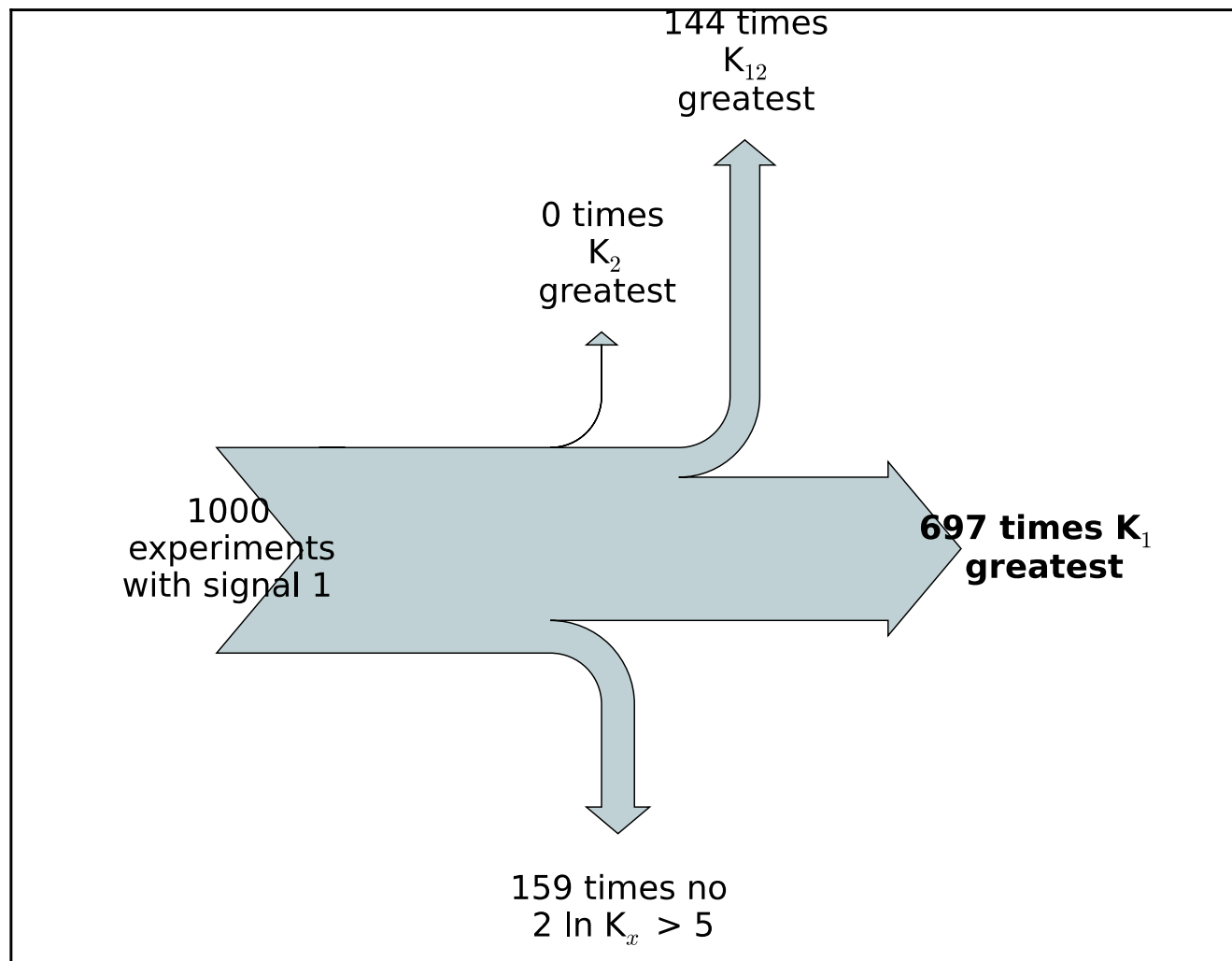
D = Data

θ = Model Parameter Vector: signal strength μ , mixing ratio r , particle masses m_j , assuming that nuisances are already dealt with

M_i = Model, e.g. T2tt or a mixture of models, e.g. T6bbWW + T2tt

(From a talk I gave about the description of signal excesses using Bayes factors)

Which signal model would be used to describe **signal 1**, feature_1 = 100 GeV?



„Pure“ signal model 1 is recognised in 70% of all cases, or in 83% of all „identifiable cases“. We never identify model 2 as the correct one.

Prelude

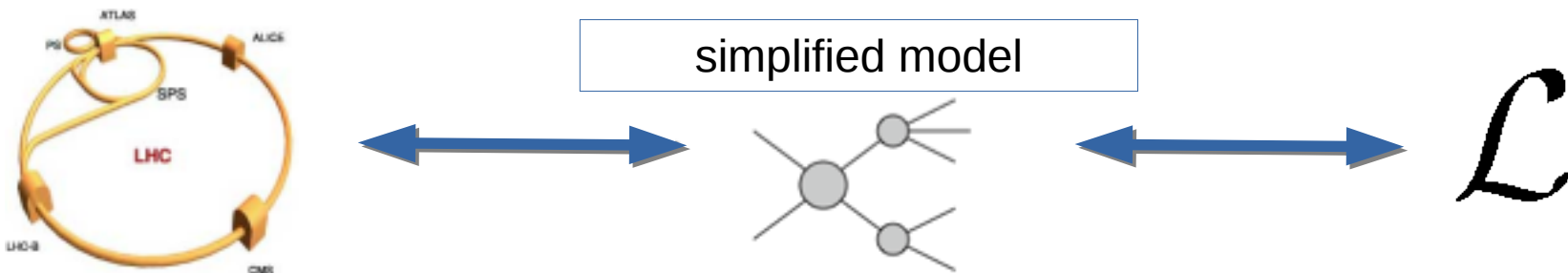
Only in a last step would I worry about “putting it all together” to infer our next, best, fundamental Lagrangian.

Top-down global fit

VS

Bottom-up incremental inference

(No doubt, as a community we want to see both approaches being followed.)



Prelude

The prelude, off-topic as it was, ends here.

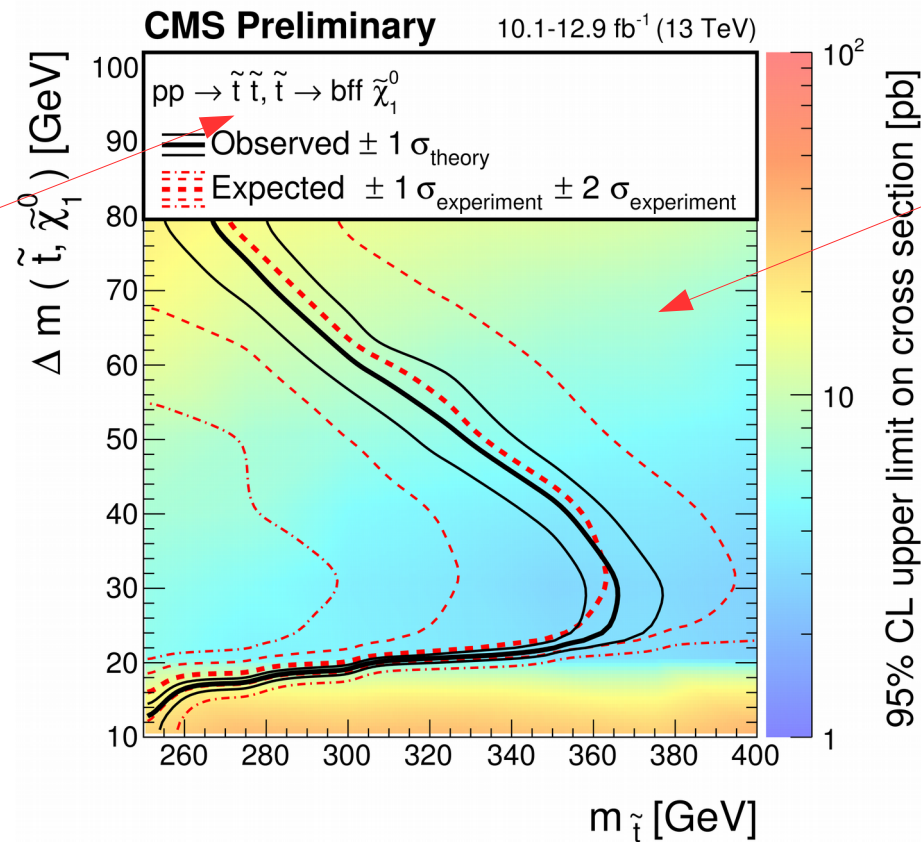
Simplified Models results



CMS-SUS-16-025: upper limits

$$\tilde{t} \tilde{t}, \tilde{t} \rightarrow b f \tilde{\chi}$$

defines what part of a full theory the result *constrains*



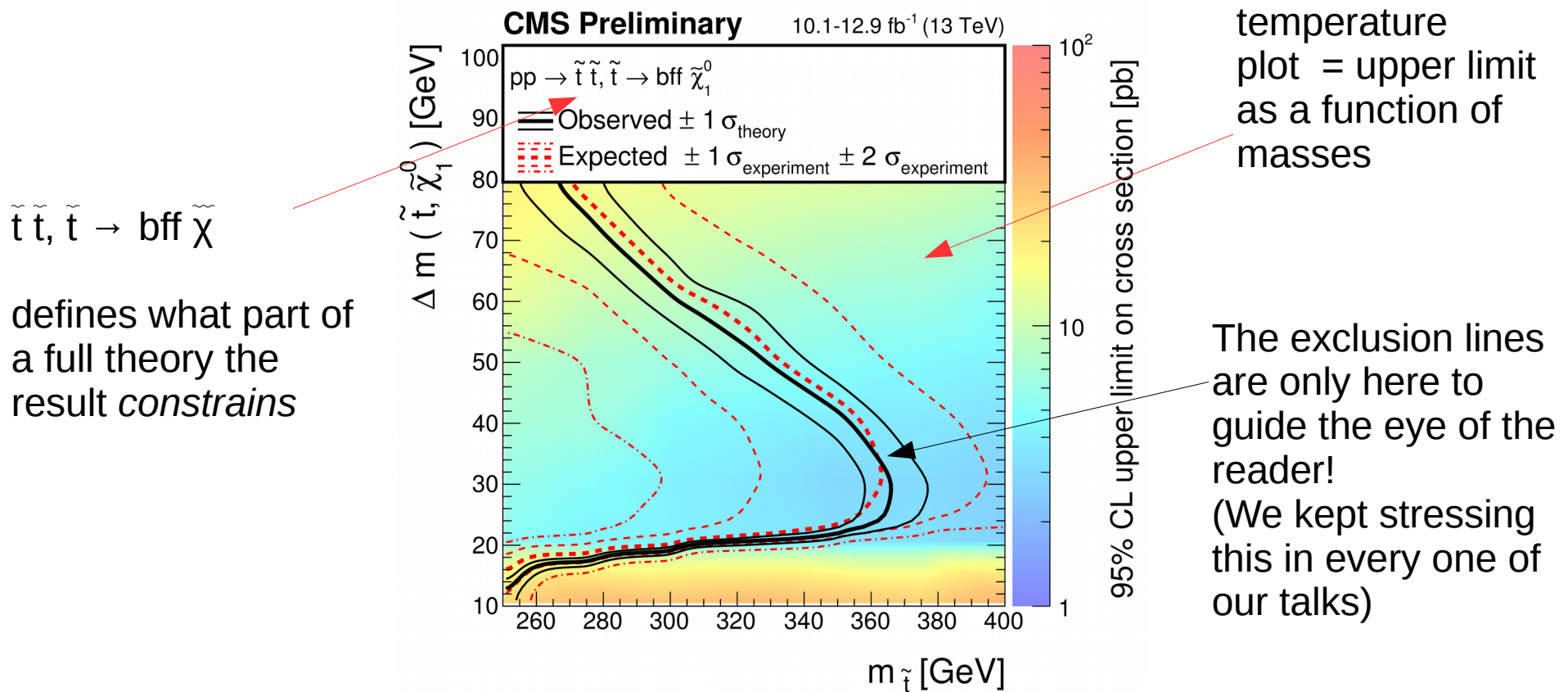
temperature plot = upper limit as a function of masses

Upper limits, parametrized in the mass space of a simplified model

Simplified Models results



CMS-SUS-16-025: upper limits

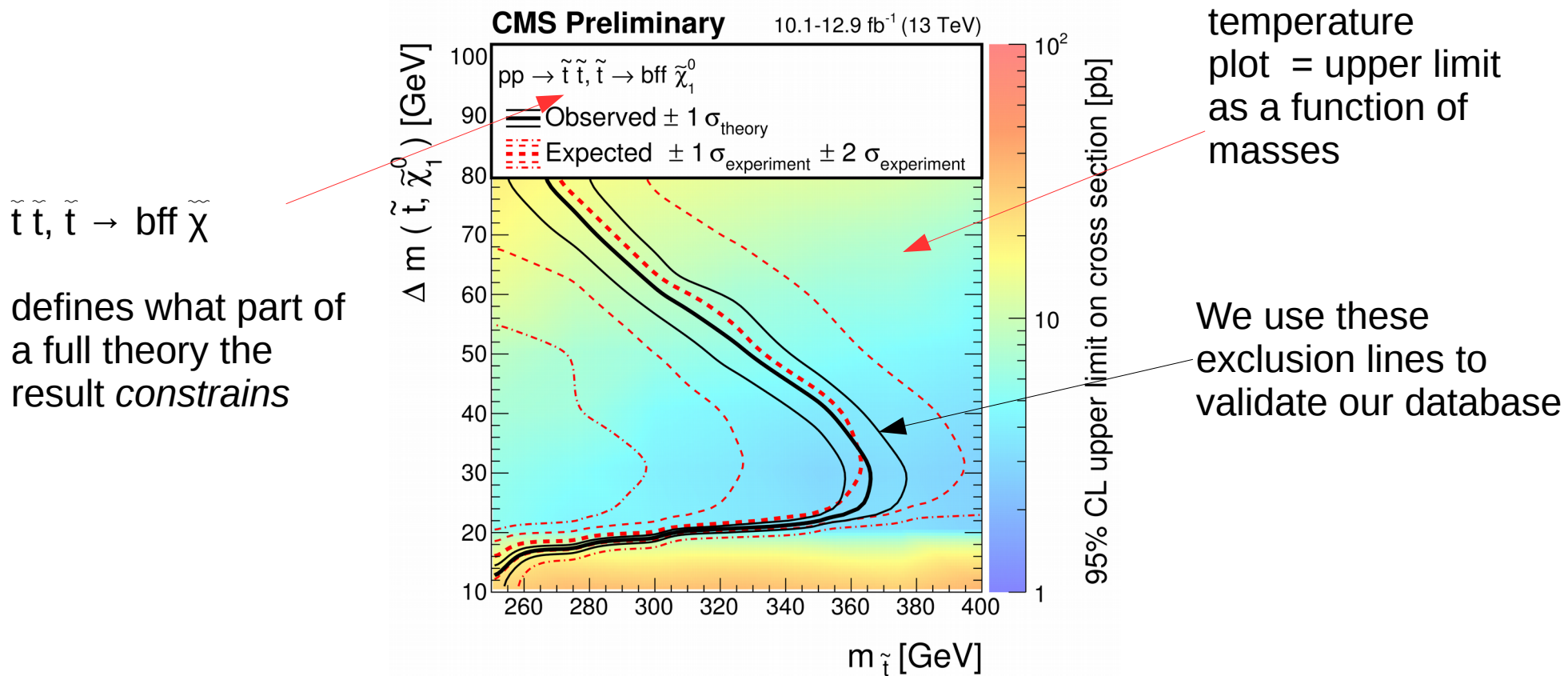


Upper limits, parametrized in the mass space of a simplified model

Simplified Models results



CMS-SUS-16-025: upper limits

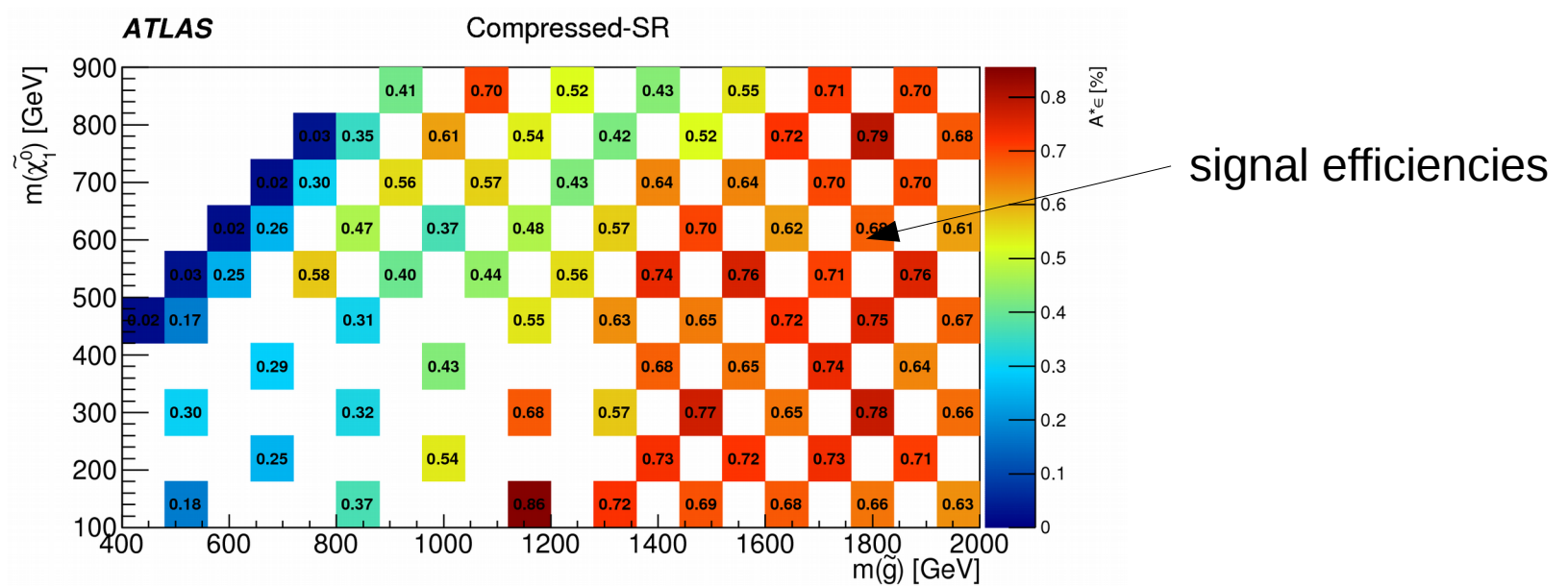


Upper limits, parametrized in the mass space of a simplified model

Simplified Models Results



ATLAS-SUSY-2016-01: efficiency maps

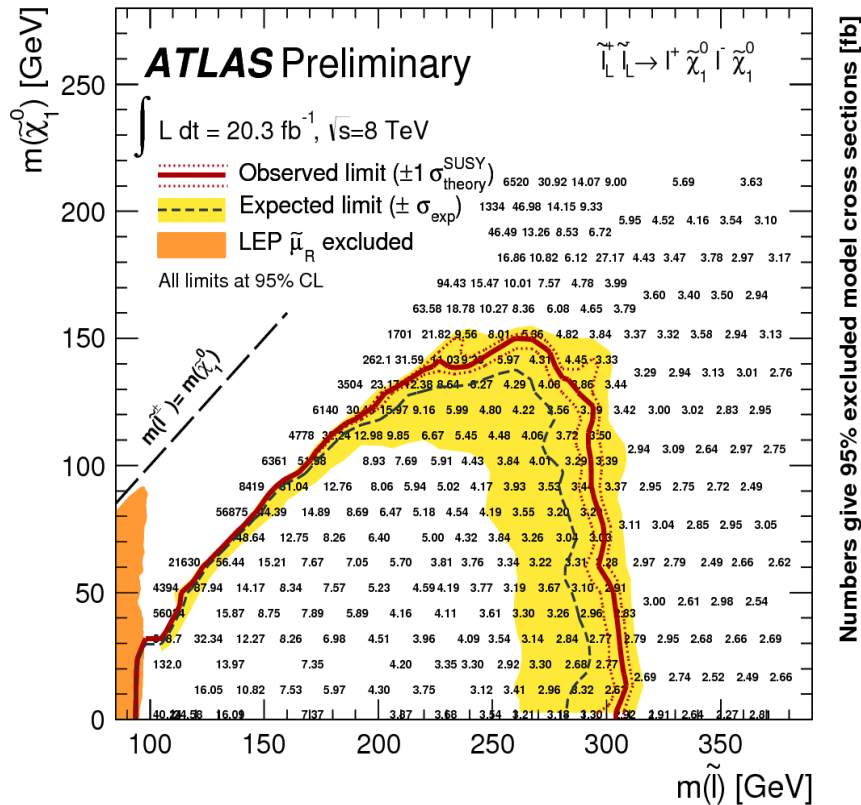


Signal efficiencies, parametrized in the mass space of a simplified model

A formal language



A formalism is needed to describe which part of a fundamental theory is constrained by what model under what conditions.



ATLAS-CONF-2013-049



Building up a database

Construction of the SModelS database



Alongside with the actual data (upper limits, efficiency maps), we keep track of All the meta information that comes with results, at the level of a specific analysis, at the level of one signal region, and at the level of one specific “result”:

```
sqrt(s): 8*TeV
lumi: 20.3/fb
id: ATLAS-CONF-2013-048
url: https://atlas.web.cern.ch/Atlas/GR...
implementedBy: WW
contact: fastlim
comment: created from fastlim-1.0
lastUpdate: 2016/08/16
```

analysis level

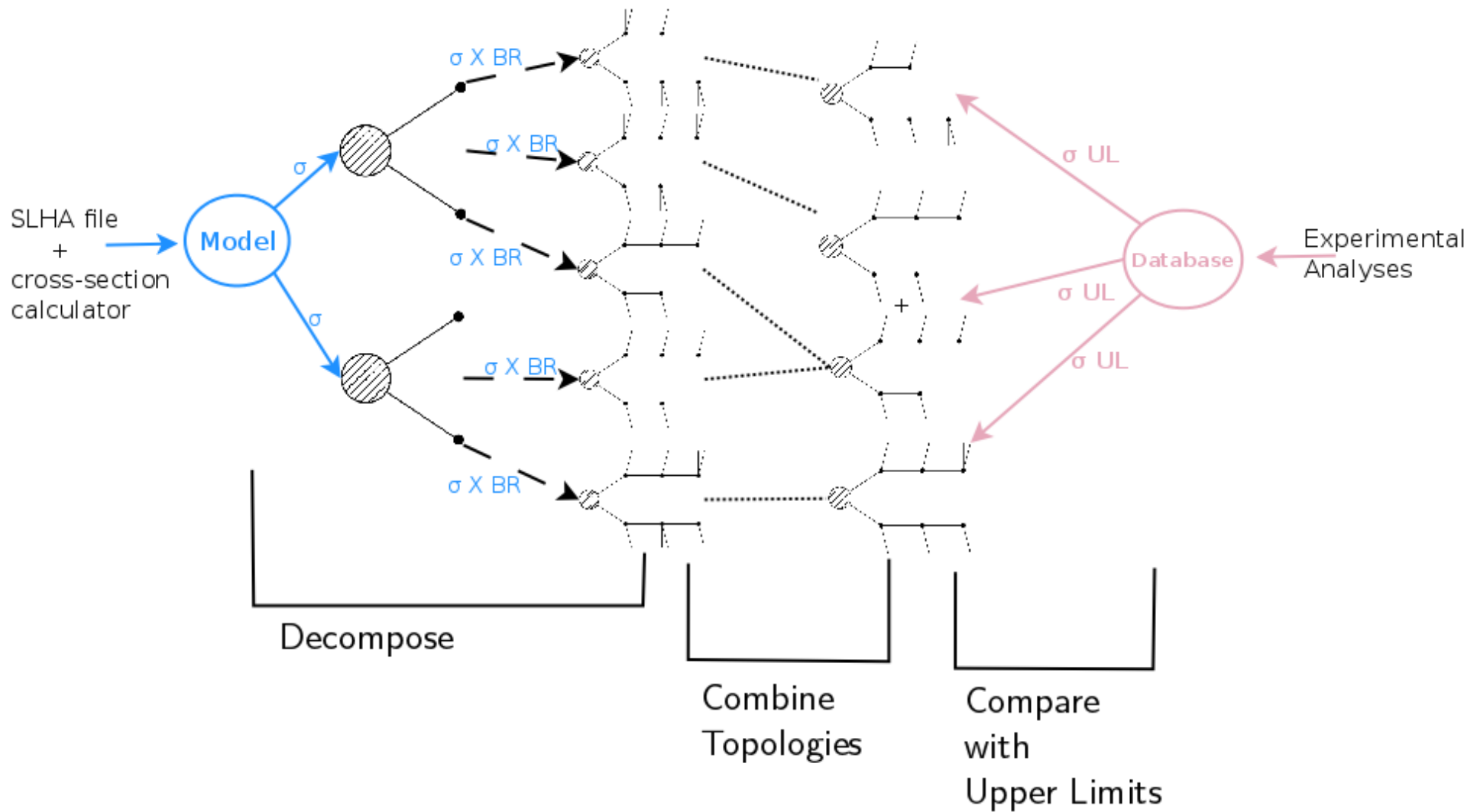
```
dataType: efficiencyMap
dataId: SR M90
observedN: 260
expectedBG: 300.0
bgError: 40.0
upperLimit: 2.5*fb
expectedUpperLimit: 4.24*fb
```

signal region

```
txName: T2
conditionDescription: None
condition: None
constraint: [[['jet']], [['jet']]]
validated: None
axes: 2*Eq(mother,x)_Eq(lsp,y)
publishedData: False
```

result

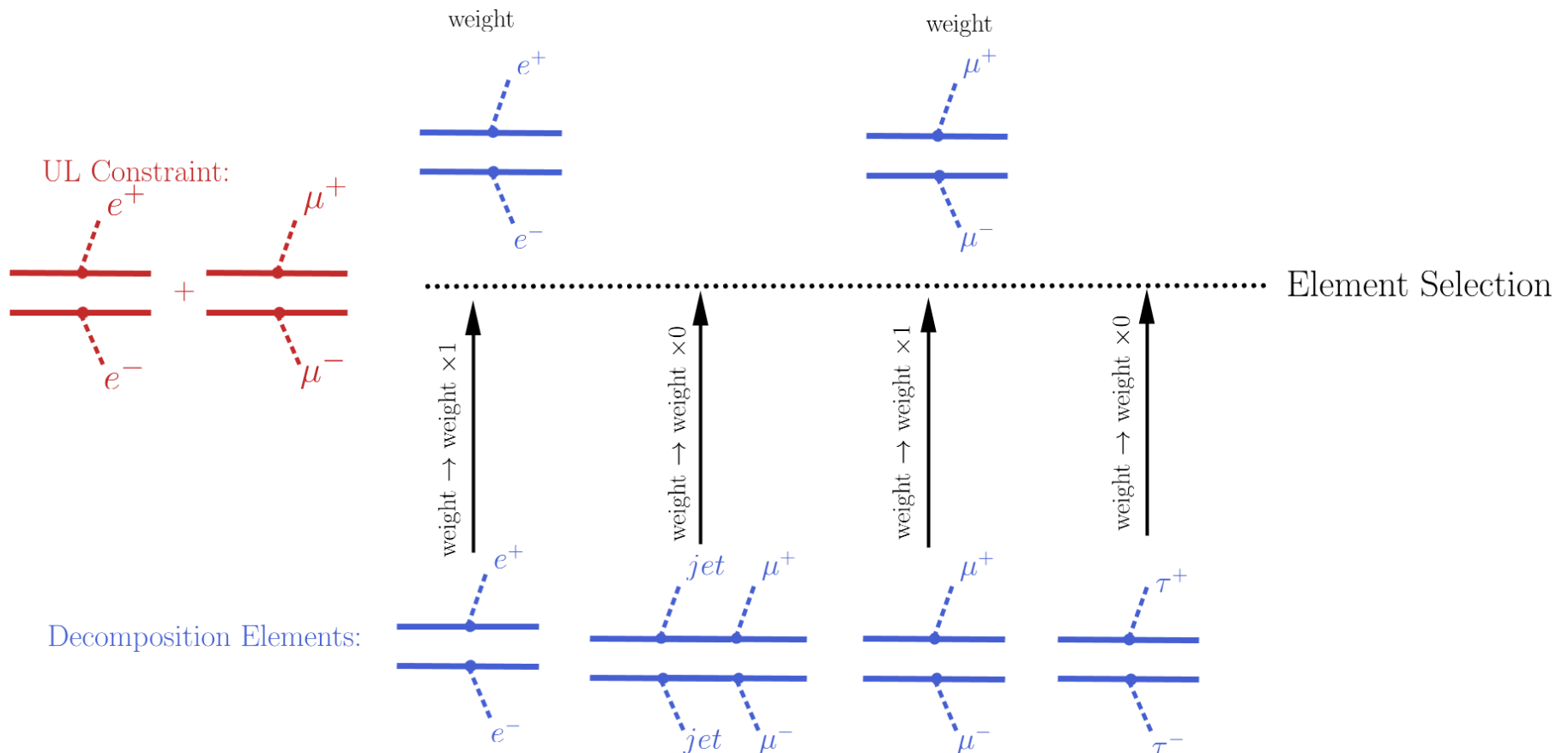
Putting it all together



Putting it all together: upper limits



For upper limit results, we cannot combine results given for different topologies. We can only apply them one-by-one. Therefore the final upper limits are very conservative. (On the positive side: it is fairly straightforward for us to add an official SMS result to our database)

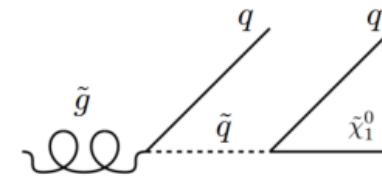
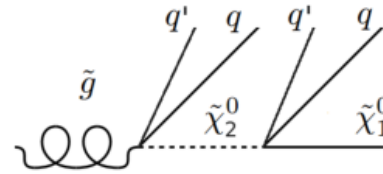
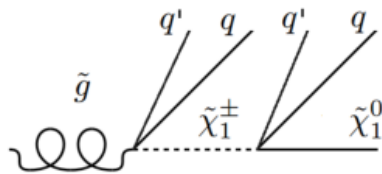


Applications: ATLAS pMSSM scan



Study Case: Missing Topologies

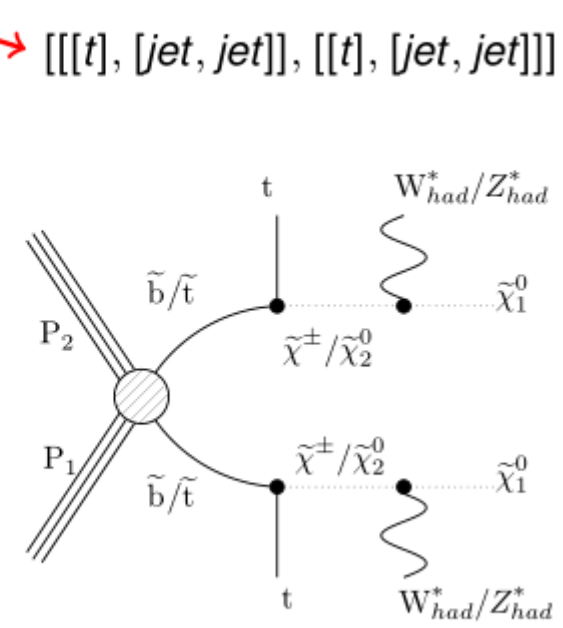
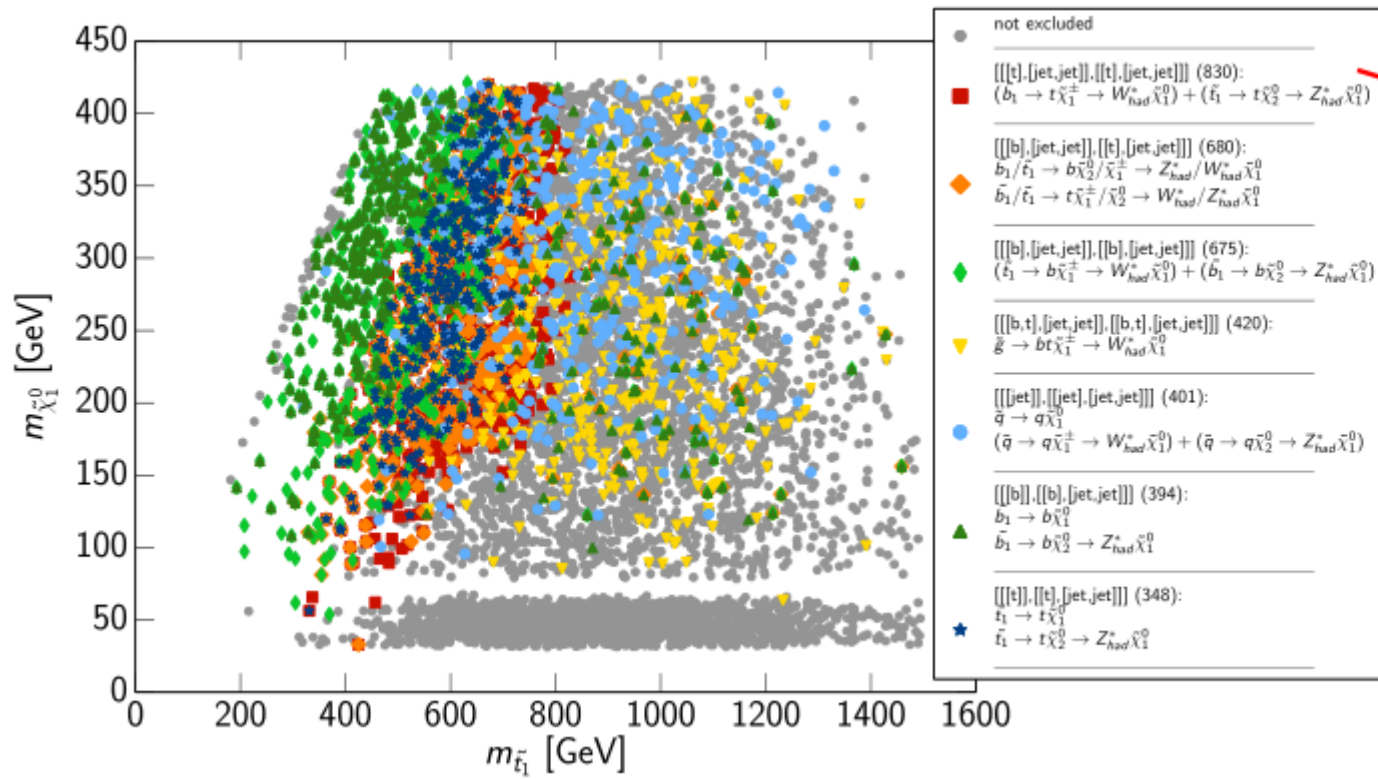
Examples of Missing Topologies



Non-degenerate right
and left handed squarks

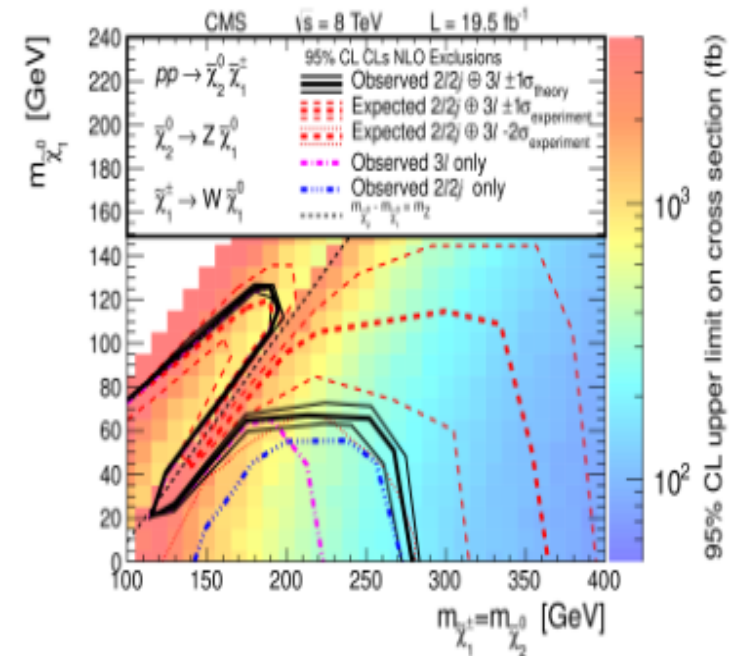
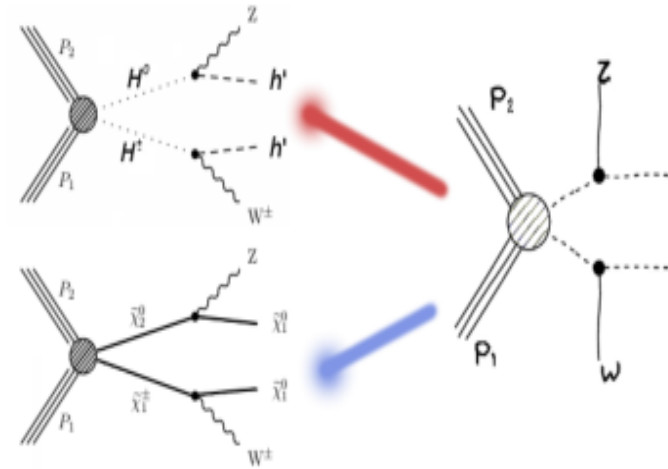
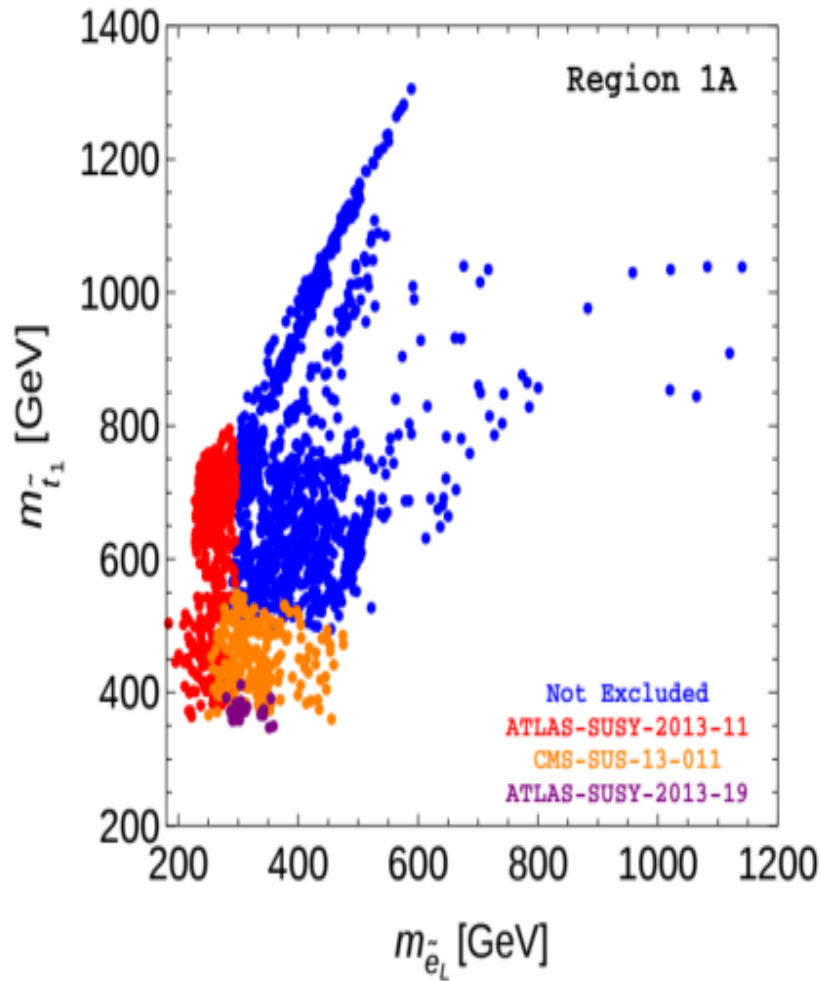
We are producing “Homegrown” efficiency maps in order to cover important missing topologies and improve SModelS constraining power

Applications: Low Fine-Tuning Scan



Which of our analyses is sensitive to this guy?

Applications: 2HDM



D. Barducci, G. Bélanger, C. Hugonie and A. Pukhov, JHEP 1601 (2016) 050

⇒ LHC constraints on 2HDM

Future

We intend to extend the functionality of SModelS in several ways:

- Extend to non- Z_2 / non-MET topologies
- Extend to long-lived particles (HCSP scenarios) and other “exotic” signatures
- Create a map of likelihoods that we can safely combine
(in good approximation: identify analyses with non-overlapping signal regions)
- Support for positive results. Prepare for the next di-photon frenzy
- Create mockup analyses that extrapolate to HL-LHC
(is quite easy for us)

<http://smodels.hephy.at>

Thank you!