

# Usage of full likelihoods in reinterpretation studies



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# Outline

- ❖ Introduction
- ❖ Other approaches: beyond pyhf
- ❖ Combining statistical models
- ❖ Computational challenges of full statistical models
- ❖ Conclusion & future directions

See Sabine's talk



See Lara's talk

# Introduction

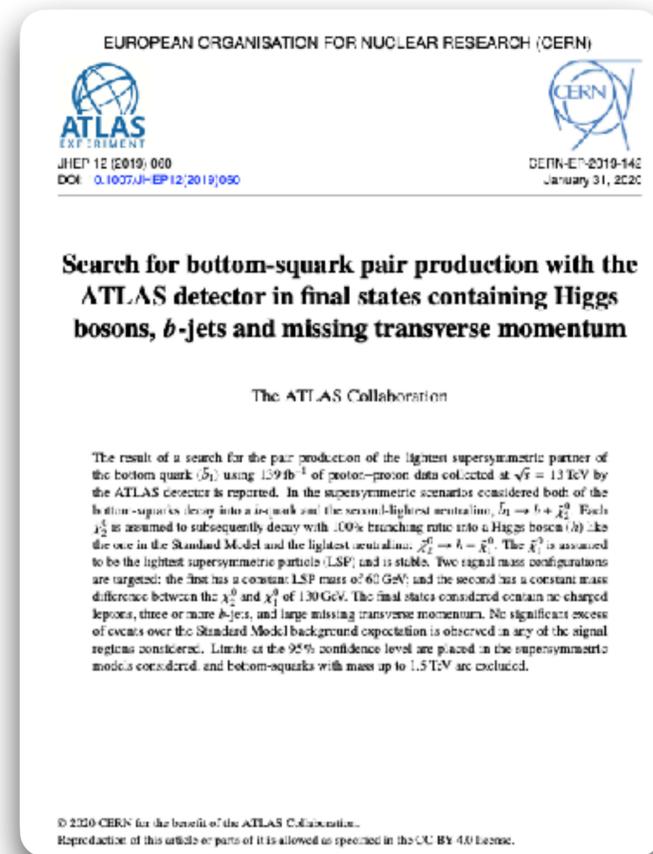
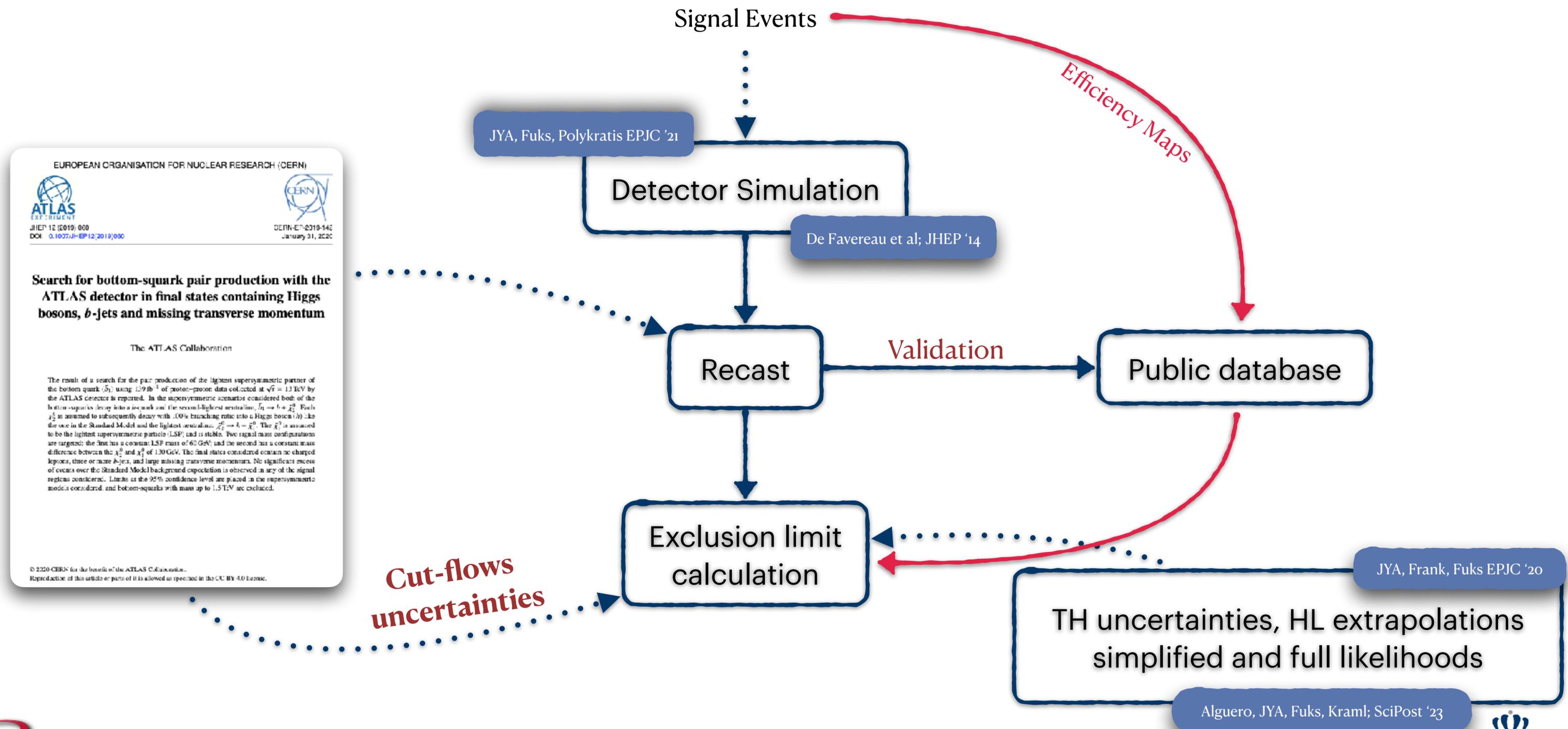
# Why designing & recasting is important?

- Exploiting the full potential of LHC (for new physics)
  - *Designing* new analyses (based on MC simulations)
  - *Recasting* LHC analyses (The LHC legacy)
- Data preservation in HEP is mandatory
  - Going beyond raw data via *analyses*
- Related tools need to be supported by the entire community
  - Both *theorists & experimentalists*
- Universal recasting tool

Les Houches Recommendations (EPJC '12)

Reinterpretation Forum Report (SciPost '20)

# (Re)interpretation of an analysis



# Statistical Models

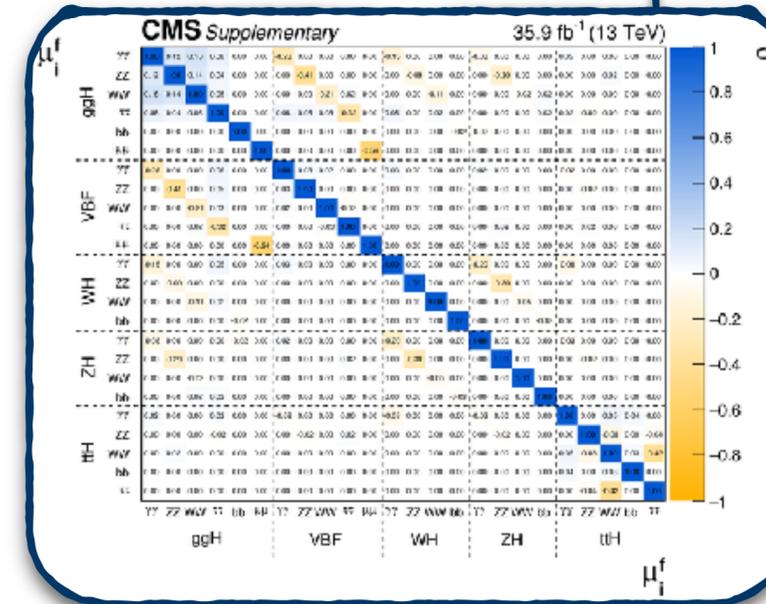
## Why important?

ATLAS SUSY and Exotics workshop  
S. Kraml '20

- The mathematical description of the analysis is provided within its statistical model.
- A likelihood enables the standard statistical approaches to extract information.
- *i.e.* how reasonably aligned the theoretical predictions with the experimental observations?

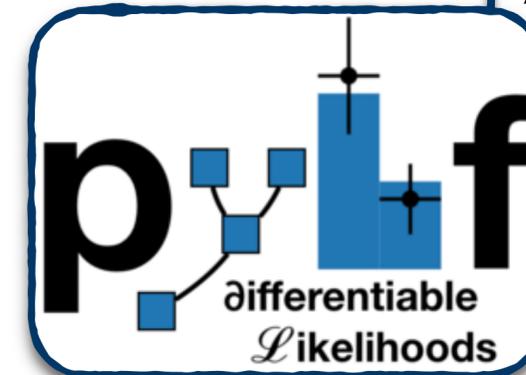
Les Houches Recommendations (EPJC '12)

Simplified likelihoods  
from CMS



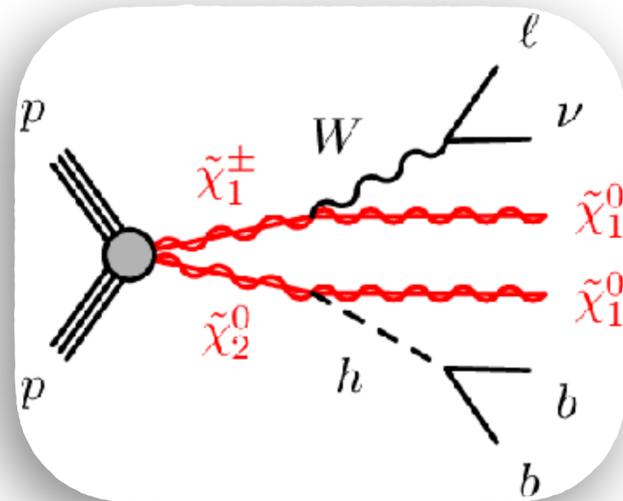
CMS-NOTE-2017-001

Full likelihoods  
from ATLAS

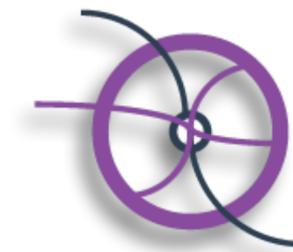


ATL-PHYS-PUB-2019-029

# Full likelihoods in action



ATLAS-SUSY-19-08



See Graeme's talk

HEPData



gz File

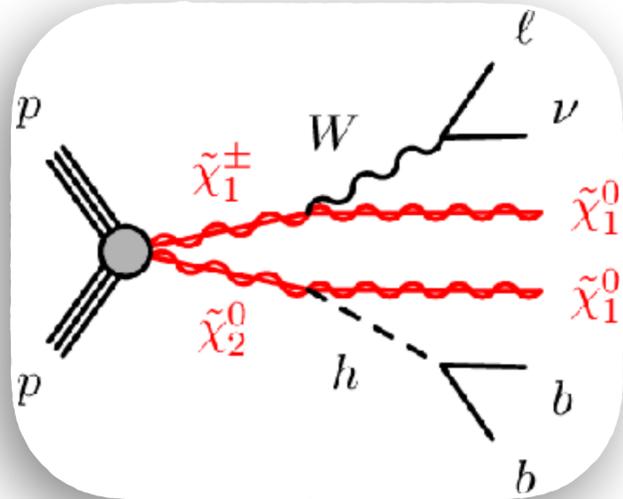
Archive of full likelihoods in the HistFactory JSON format described in CERN-EP-2019-188. For each signal point the background-only model is found in the file named BkgOnly.json. All jsonpatches are contained in the file patchset.json. Each patch is identified in patchset.json by the metadata field "name": "C1N2\_Wh\_hbb\_[m1]\_[m2]" where m1 is the mass of both the lightest chargino and the next-to-lightest neutralino (which are assumed to be nearly mass degenerate) and m2 is the mass of the lightest neutralino.

Download

- ❖ Wino-like electroweakinos
- ❖ ATLAS shares **HistFactory like** json files to form full profile likelihoods.
- ❖ Each file includes detailed information on backgrounds and corresponding nuisance parameters.

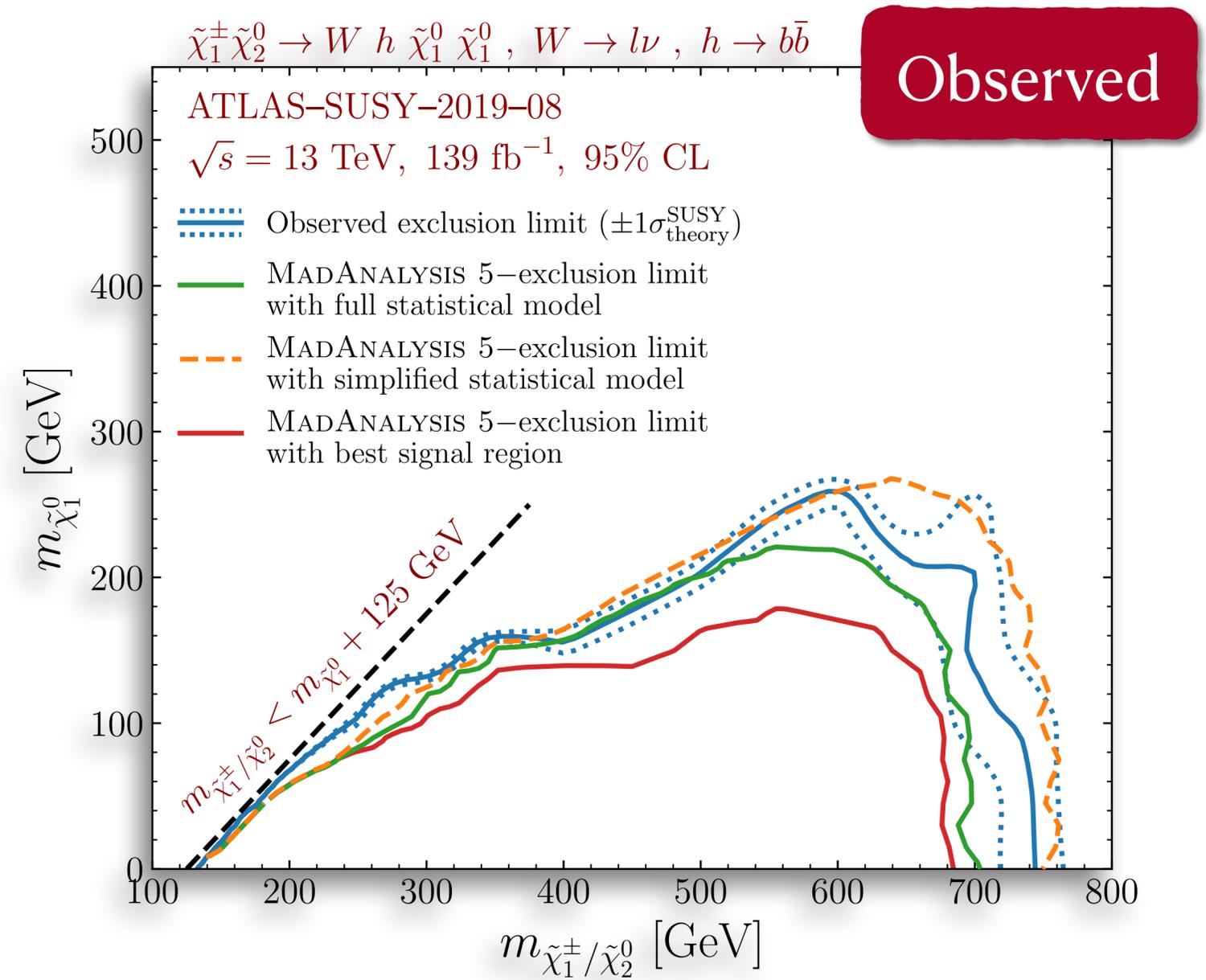
# Full likelihoods in action

Alguero, JYA, Fuks, Kraml; SciPost '23



- ❖ Wino-like electroweakinos
- ❖ ATLAS shares **HistFactory-like** JSON files to form full profile likelihoods.
- ❖ **The difference between green and blue is due to the lack of CR/VR!**
- ❖ Simplified likelihoods are achieved by compressing all the information into a **single nuisance parameter**.

 [eschanet/simplify](https://github.com/eschanet/simplify)



# What is the next step?

- ❖ Not all analyses come with “pyhf” likelihoods (I wish!)
- ❖ Full likelihoods can become computationally intensive; can we simplify them without sacrifice?
- ❖ There are many different software for hypothesis testing; how can we unite them?
- ❖ How can we combine statistical models?

(Re)interpretation of the LHC results for new physics

29 August 2023 to 1 September  
Durham University

Also see Carsten's  
talk from yesterday

HS<sup>3</sup>

High Energy Physics



Statistics Serialization Standard

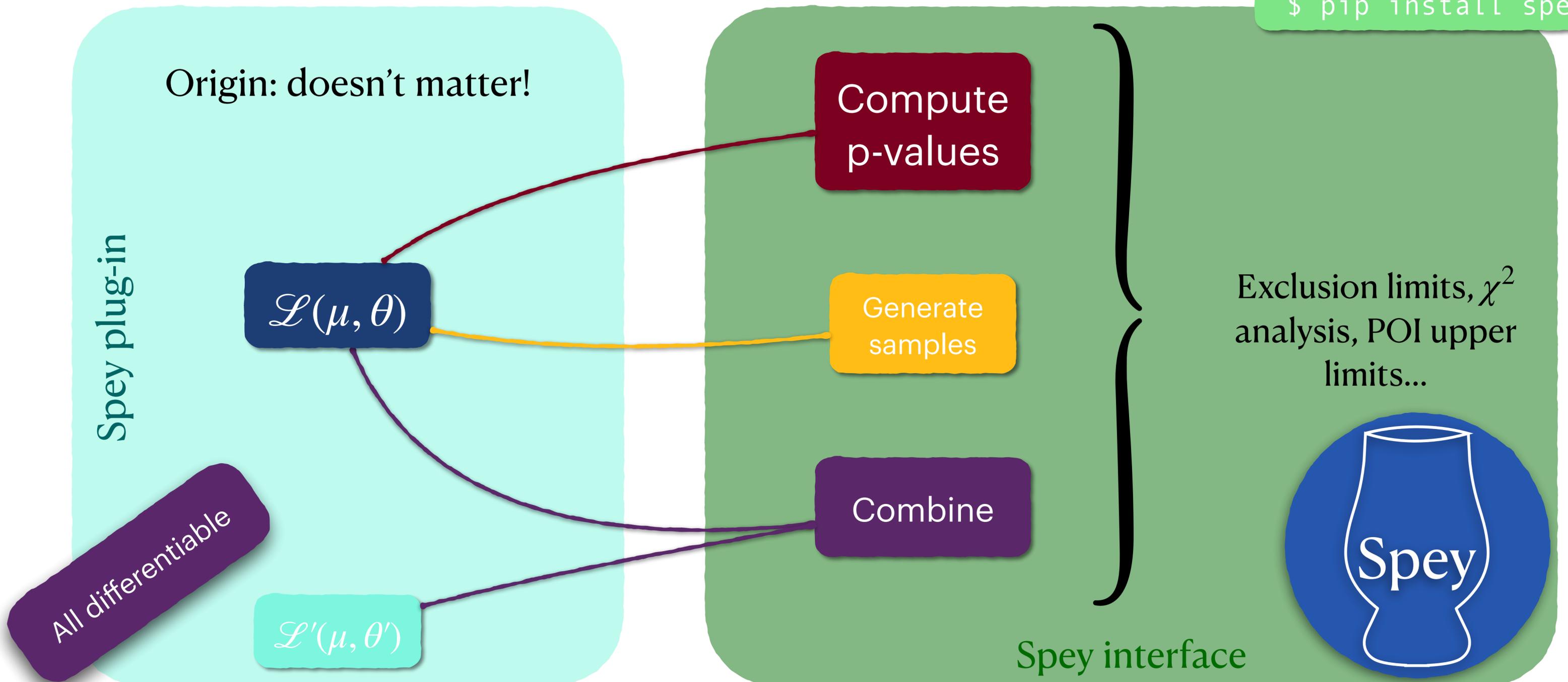
Carsten Burgard

Tomas Dado, Jonas Eschle, Matthew Feickert, Cornelius Grunwald,  
Alexander Held, Robin Pelkner, Jonas Rembser, Oliver Schulz

tu technische universität  
dortmund

# Uniting statistical modelling under Spey

\$ pip install spey

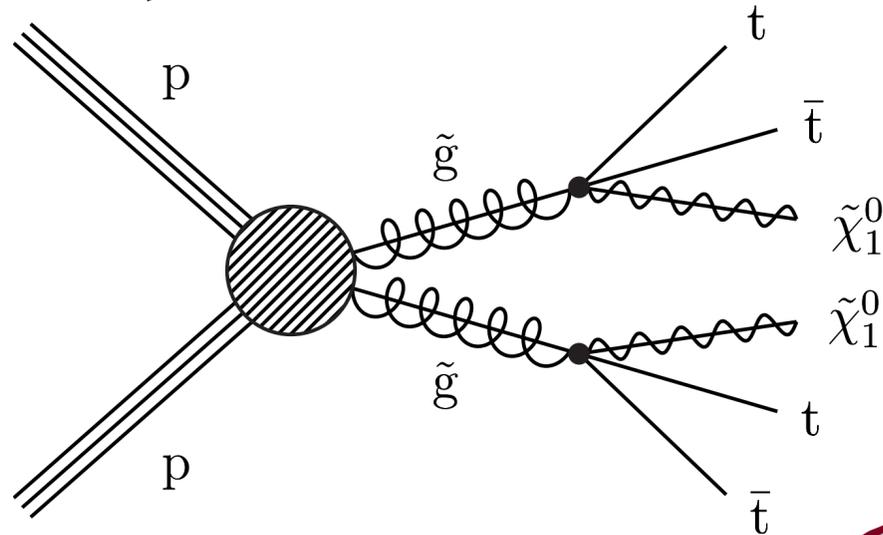


# Simplified Likelihoods

Not all analyses come with “pyhf” likelihoods

# Simplified likelihoods

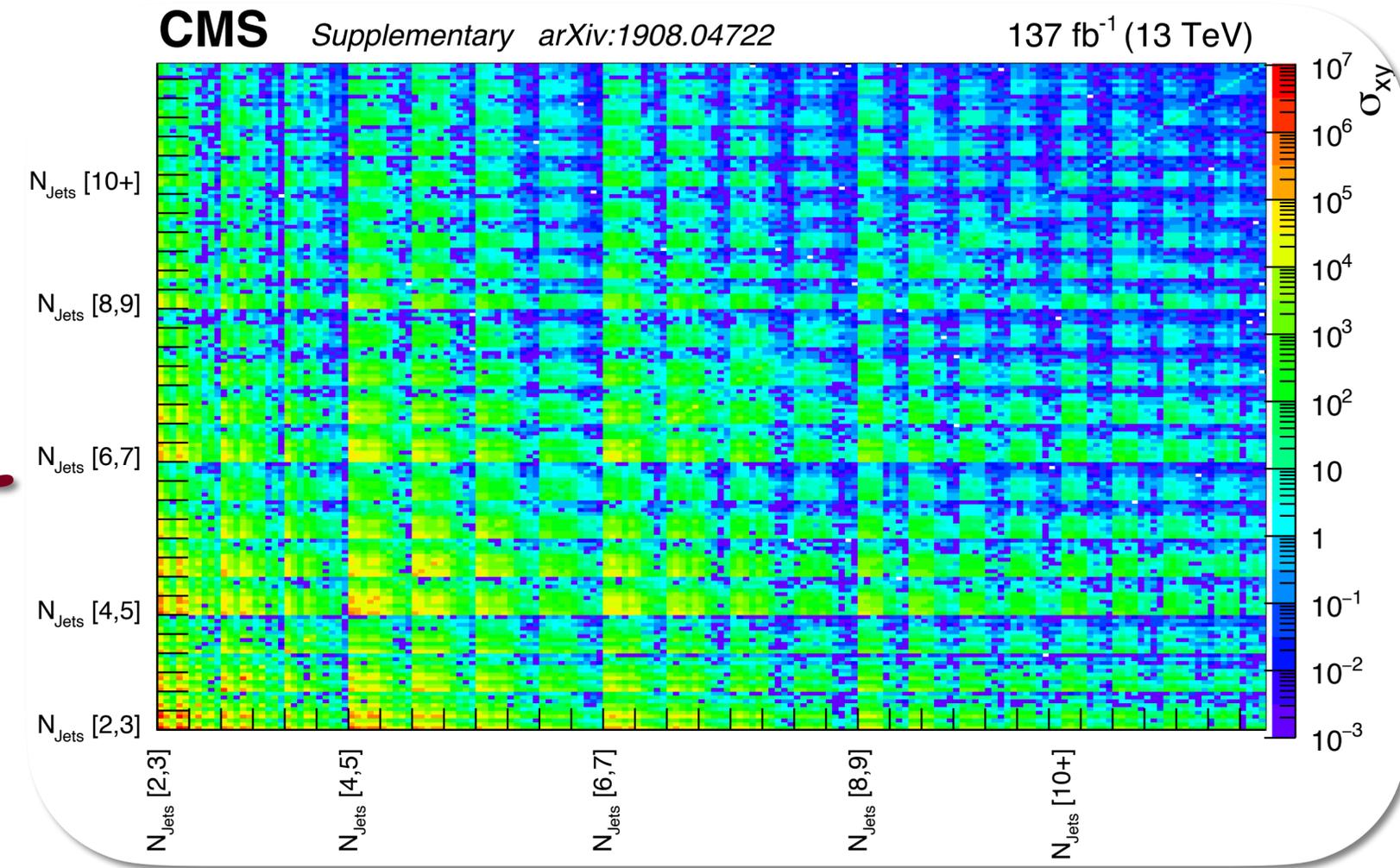
CMS-SUS-19-006



$$\mathcal{L}(\mu, \theta) = \left[ \prod_{i \in \text{bins}} \text{Pois} (n^i | \mu n_s^i + n_b^i + \theta^i \sigma_b^i) \right] \cdot \mathcal{N}(\theta | 0, \rho)$$

CMS-NOTE-2017-001

Can be expanded via third moments or using asymmetric uncertainties

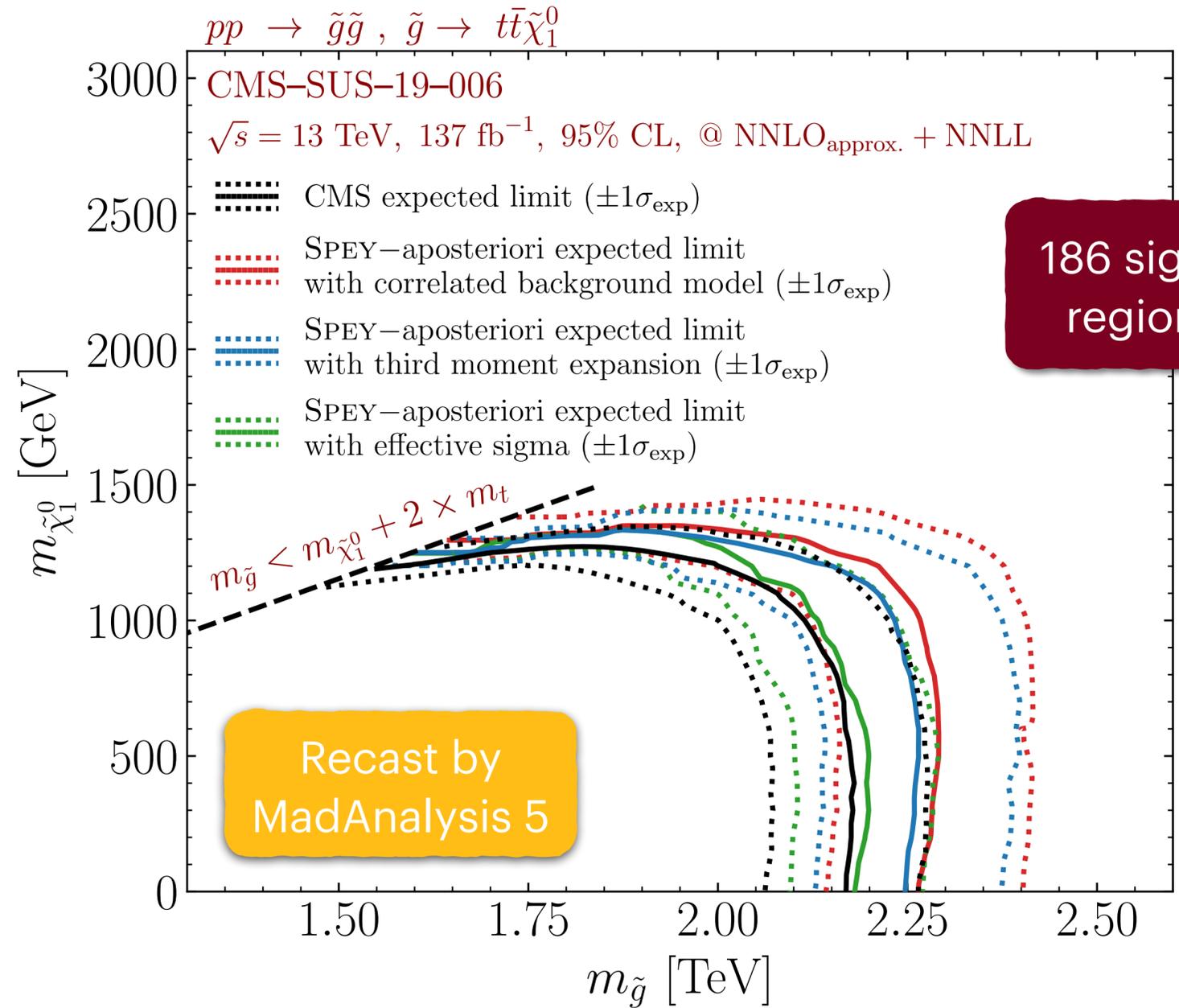
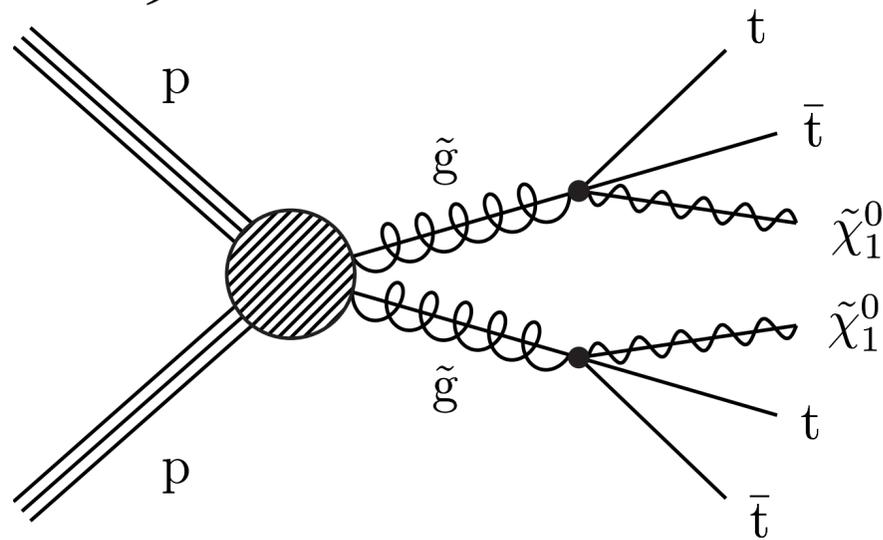


186 signal regions!

# Simplified likelihoods

JYA; arXiv:2307.06996

CMS-SUS-19-006

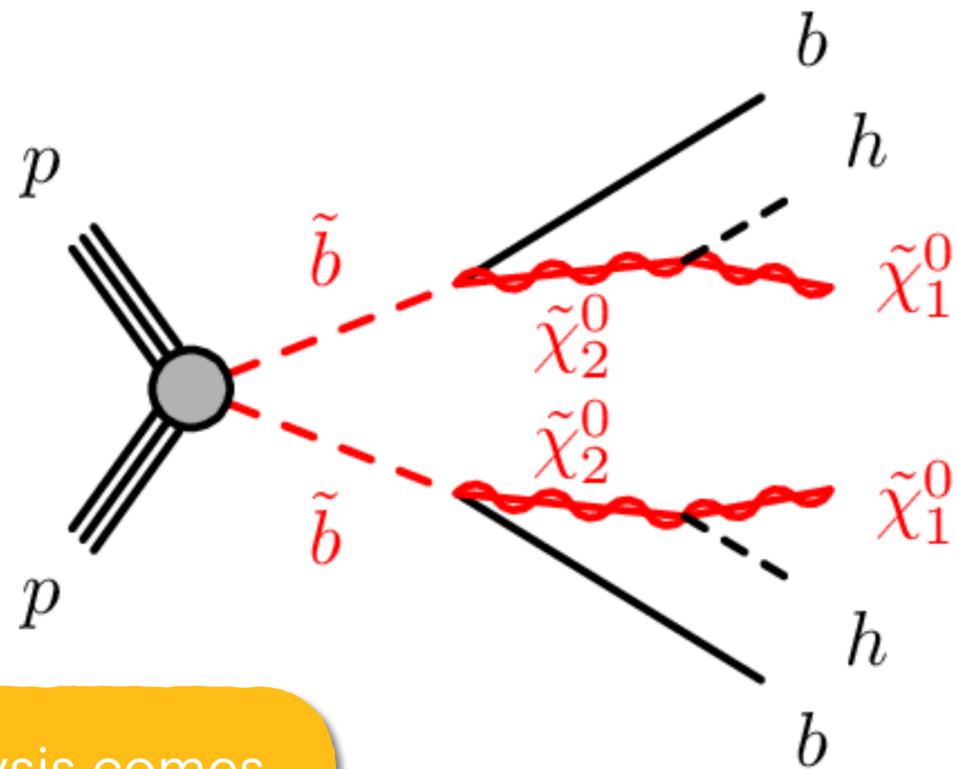


# Combining Statistical Models

# Combining Statistical Models

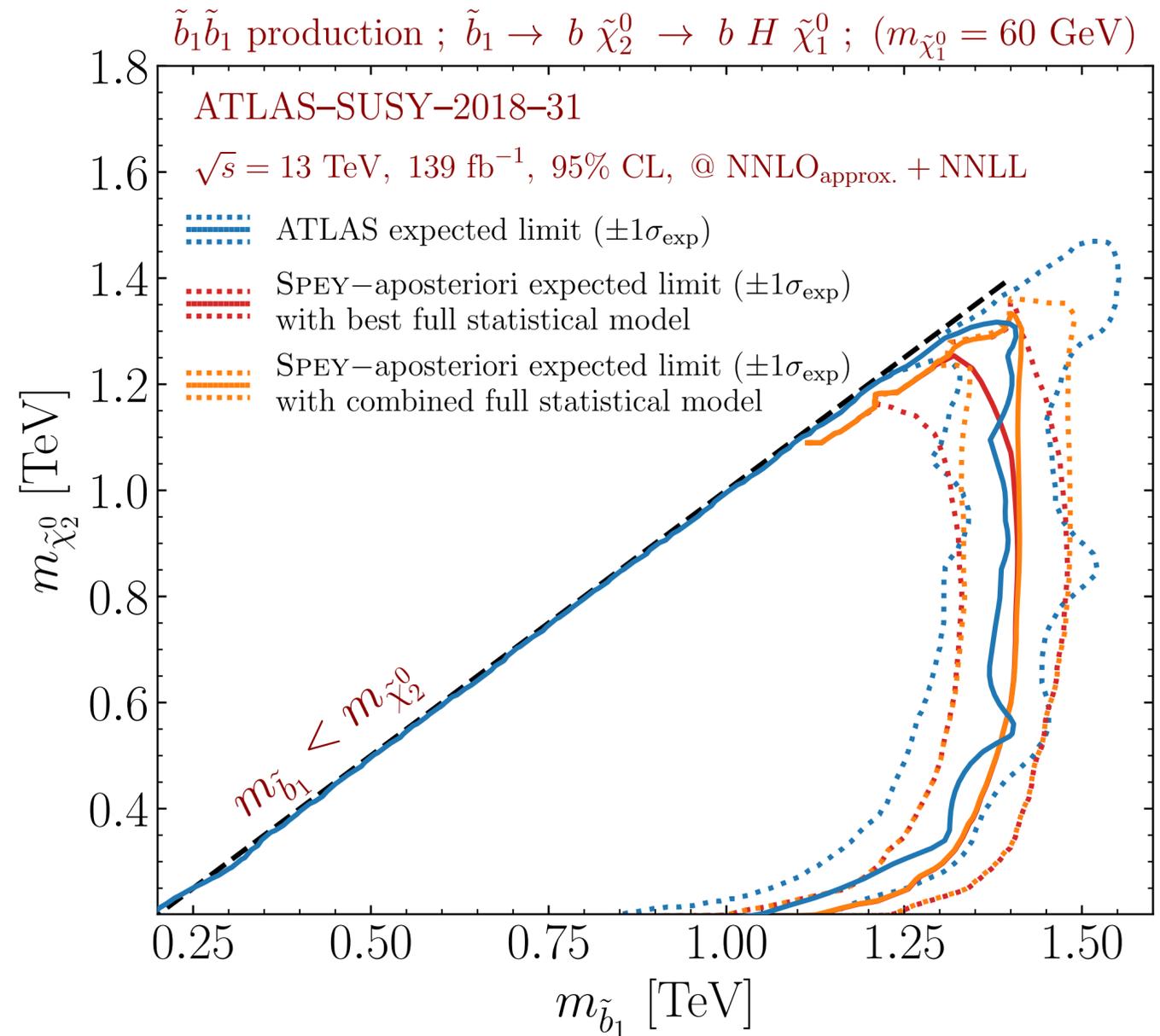
\$ pip install spey-pyhf

JYA; arXiv:2307.06996



This analysis comes with three different super regions!

Full likelihoods include all the necessary information to mix and match nuisance parameters to combine them!



Recast by MadAnalysis 5

Jack Y. Araz

# Combining Statistical Models

JYA; arXiv:2307.06996

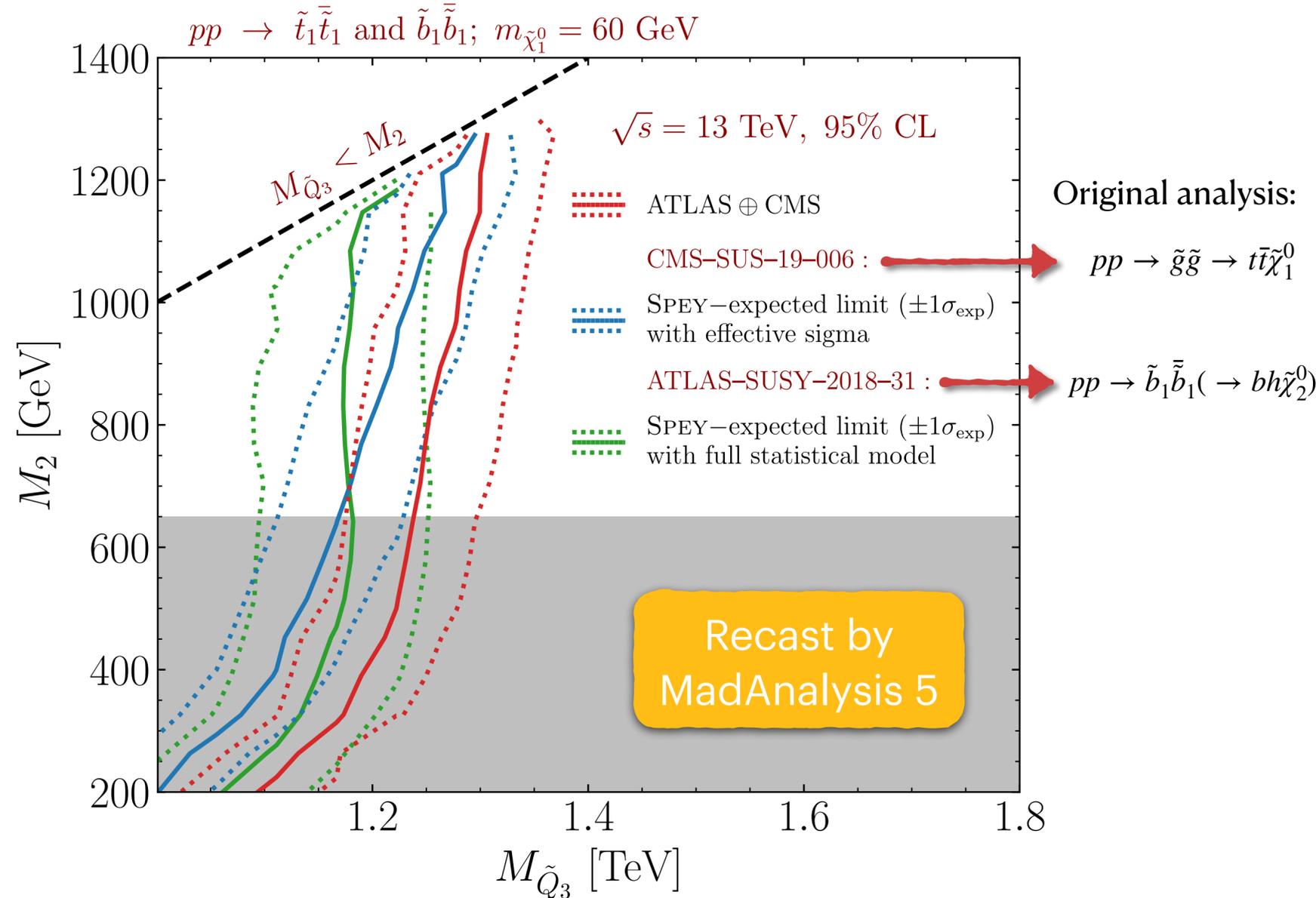
$$\mathcal{L}' = \mathcal{L}_{\text{ATLAS}} \oplus \mathcal{L}_{\text{CMS}}$$

Full likelihood  $\rightarrow$   $\mathcal{L}_{\text{ATLAS}}$       Simplified likelihood with effective sigma model  $\rightarrow$   $\mathcal{L}_{\text{CMS}}$

A combination of analyses, rather than regions, contains much more information!



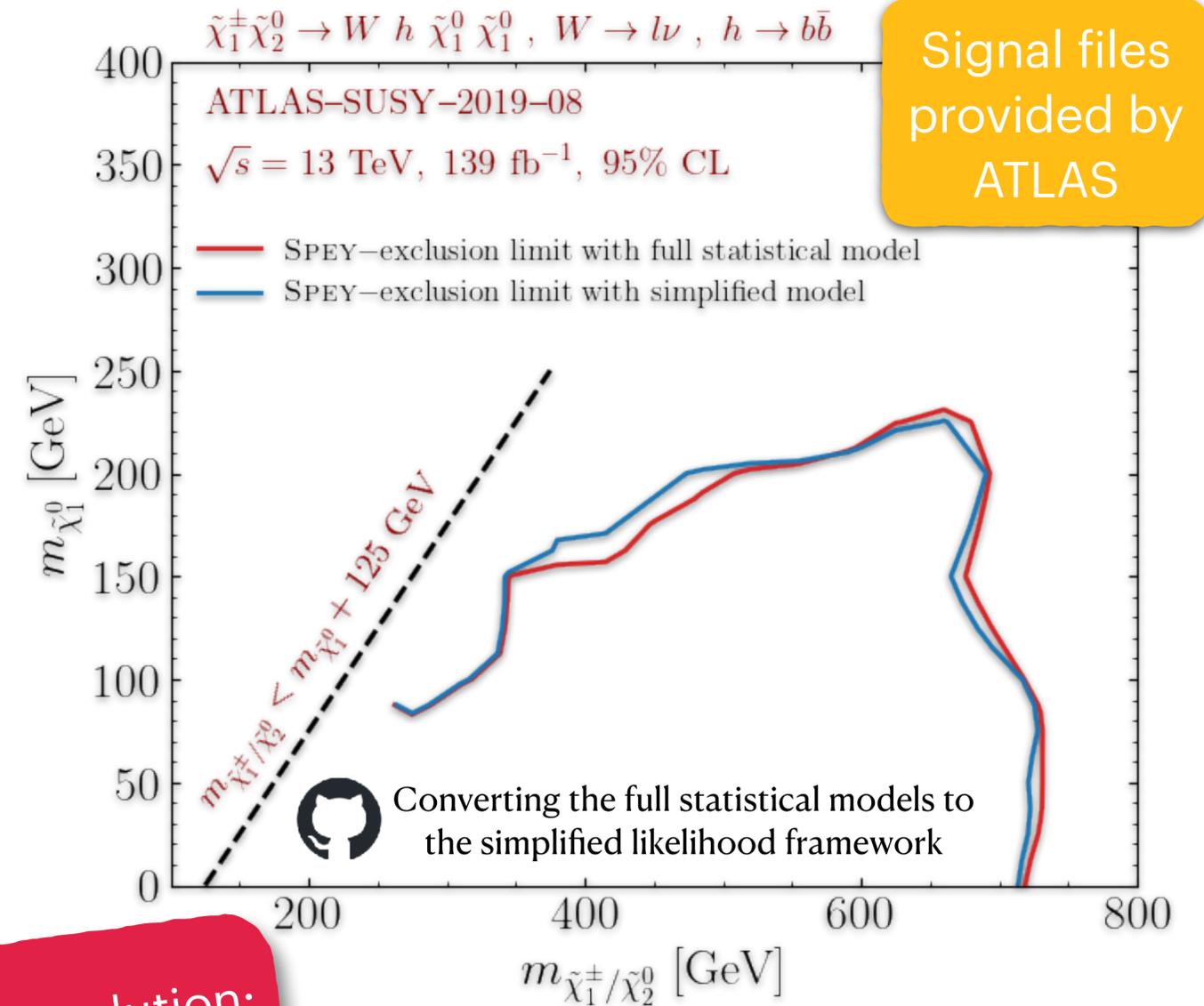
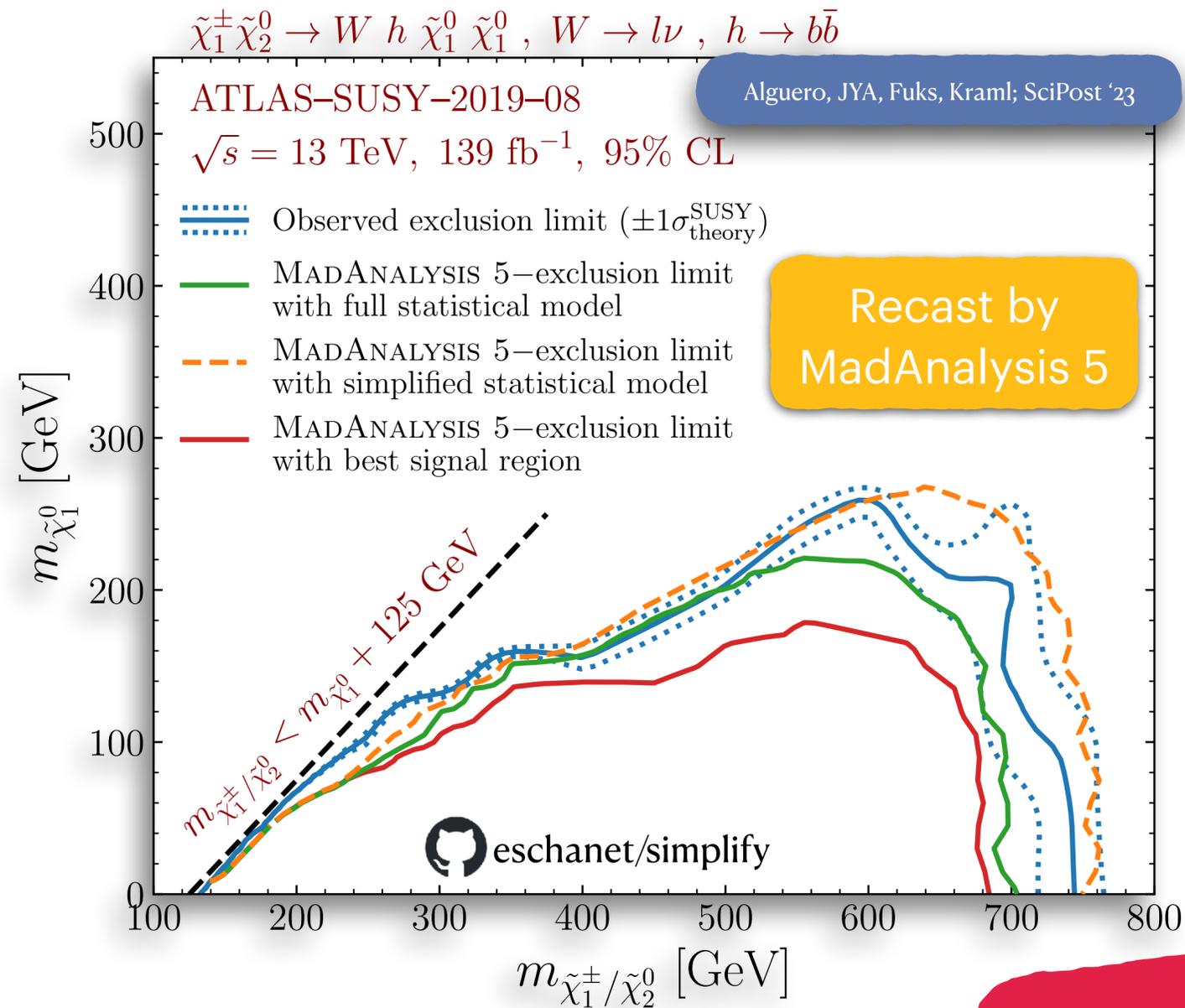
Assumes that likelihoods are not correlated



**How do we avoid the  
computational cost of full  
likelihoods?**

# Faster inference!

JYA; arXiv:2307.06996



The next step in the evolution:  
 Machine-learned likelihoods  
 with Humberto & Wolfgang

# Conclusion & Future Directions

# Conclusion



- ❖ CMS planning to release full likelihoods via the “combine” package
- ❖ Spey acts as a hub for various likelihood implementations for hypothesis testing.
- ❖ The ability to study likelihoods in a backend-agnostic way opens up various possibilities, such as combinations.
- ❖ There are other statistical model implementations!  [fastprof-hep/fastprof](https://github.com/fastprof-hep/fastprof)

Smarter simplified likelihood implementations

Machine Learned likelihoods

Dedicated custom likelihoods for other experiments

Thanks to pyhf-dev team, who pushed all this endeavour!