



# UC SANTA CRUZ



[mapyde](https://mapyde.com)

# Reduce, Reuse, Reinterpret

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and Prof. Mike Hance, Camila Ots  
August 29th, 2023

[indico.dur.ac.uk://event/1178/](https://indico.dur.ac.uk://event/1178/)

[arXiv:2306.11055](https://arxiv.org/abs/2306.11055)




Event: 2418777995

2016-06-04 03:47:03 1E

# Overview of Today

How to reuse a public ATLAS analysis

- ✦ The **typical workflow** for published analyses
- ✦ Reinterpretation of **ATLAS EWK Compressed**
- ✦ **Technical details** of mapyde 
- ✦ **Implementing** custom workflows

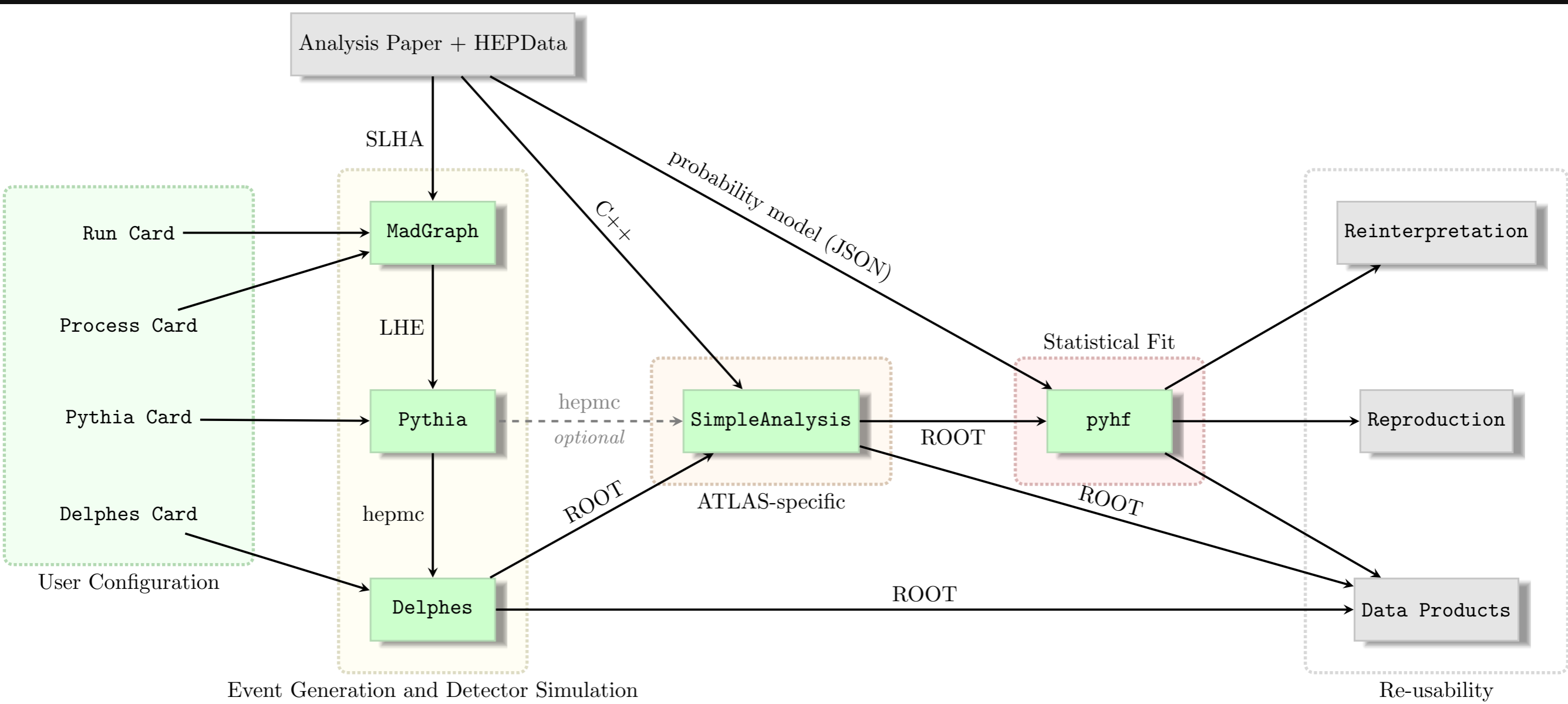
 **Let's get started**

# What is mapyde?

- ✦ MaPyDe: a highly configurable pipelining framework for recycling public analyses
  - ✦ named after common event generation and simulation toolkits: MADGRAPH, PYTHIA8, and DELPHES
- ✦ SW dependencies: python, docker/singularity/apptainer
- ✦ Additional tools exist (and new ones can be supported)
  - ✦ SimpleAnalysis: [ATL-PHYS-PUB-2022-017](#)
  - ✦ pyhf: [doi://10.21105/joss.02823](#)

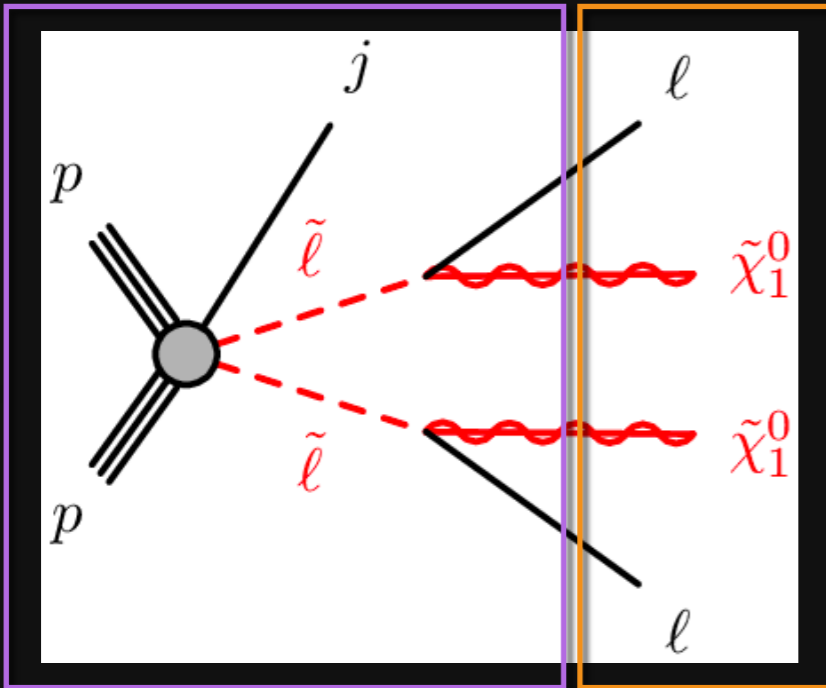
**i See my [tutorial last year at RiF2022](#)**  
*approachable pheno tooling for students(!)*

# Workflow



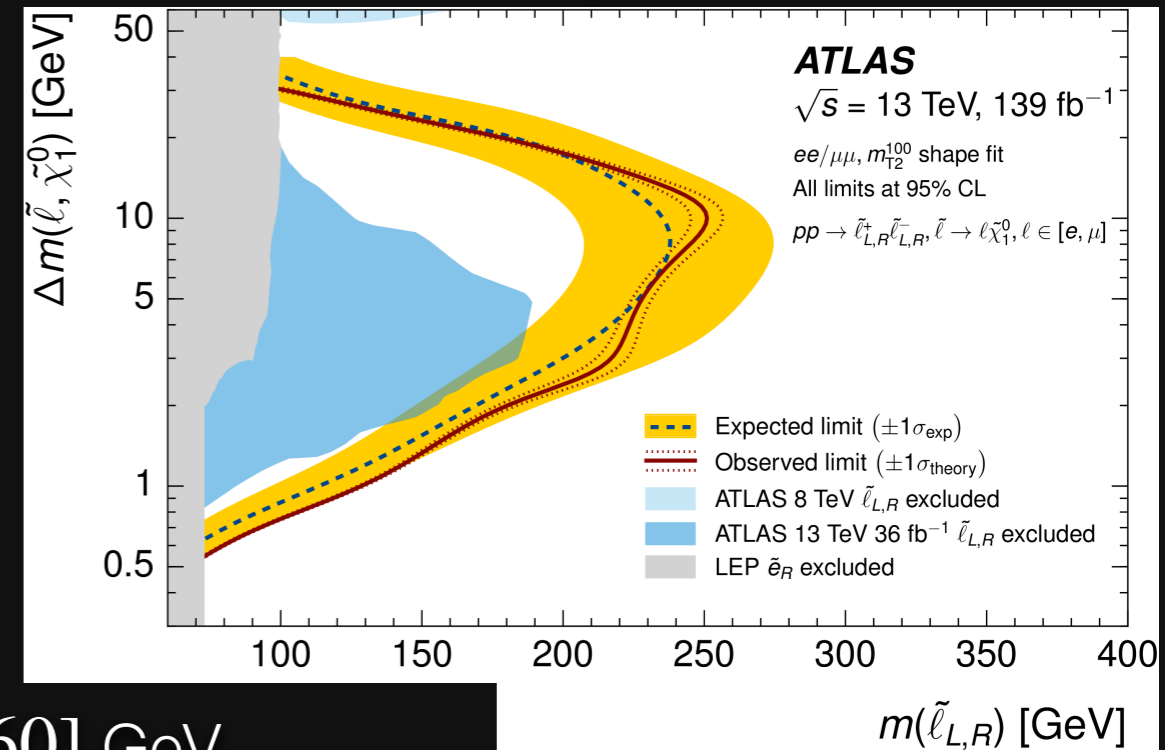
# A reminder...

sleptons

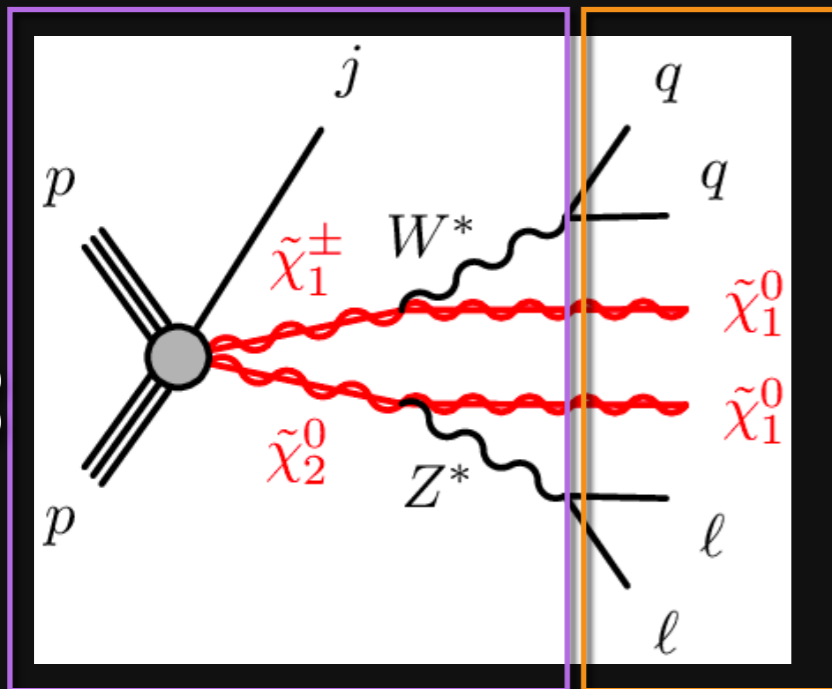


$$\Delta m(\tilde{\chi}_2^0, \tilde{\chi}_1^0) \in [1, 60] \text{ GeV}$$

signature:  $2\ell + 3j + E_T^{\text{miss}}$

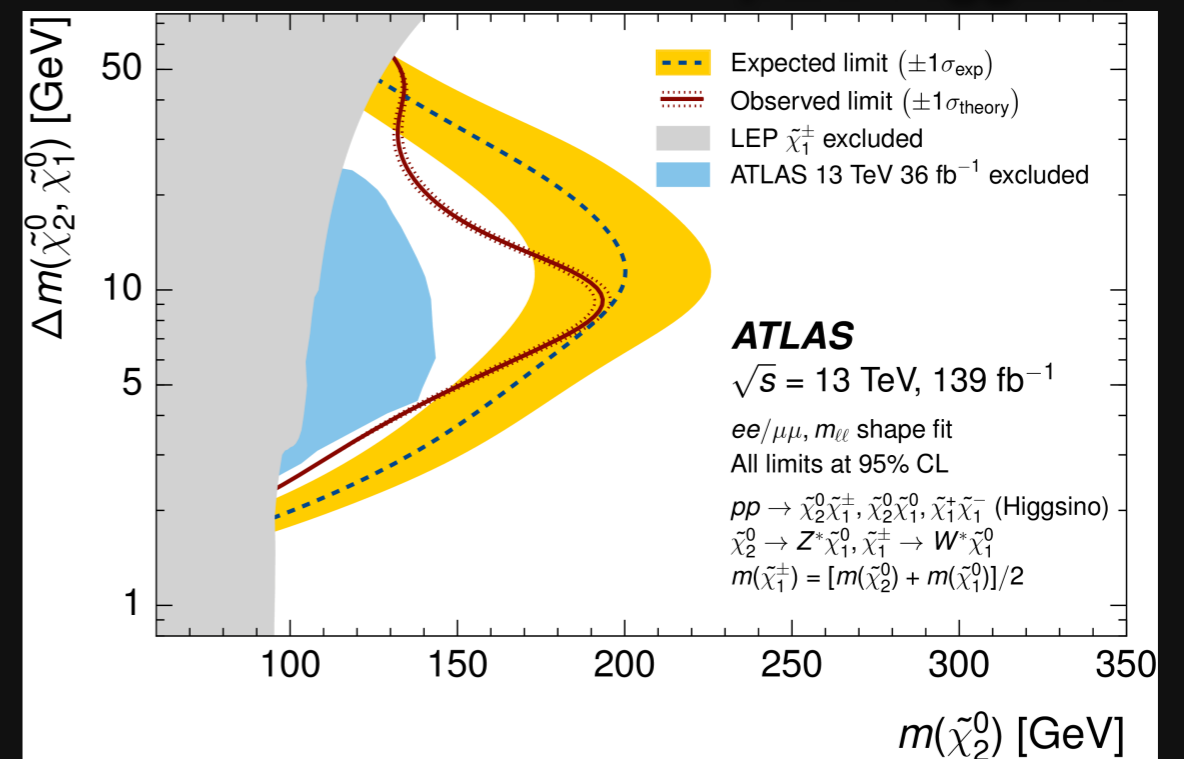


Higgsinos



signature

“pure-Higgsino”



model

Let's use new models and reinterpret!

# Public data

HEPData Search HEPData Search

Searches for electroweak production of supersymmetric particles with compressed mass spectra in  $\sqrt{s} = 13$  TeV  $pp$  collisions with the ATLAS detector

The ATLAS collaboration  
Aad, Georges, Abbott, Brad, Abbott, Dale Charles, Abed Abud, Adam, Abelling, Kira, Abhayasingh, Deshan Kavishka, Abidi, Syed Haider, Abouzeid, Ossama, Abraham, Nicola, Abramowicz, Halina  
Phys.Rev.D 101 (2020) 052005, 2020.  
<https://doi.org/10.17182/hepdata.91374.v5>

Journal INSPIRE Resources

Abstract (data abstract)  
This paper presents results of searches for electroweak production of supersymmetric particles in models with compressed mass spectra. The searches use 139/fb of  $\sqrt{s}=13$  TeV proton-proton collision data collected by the ATLAS experiment at the Large Hadron Collider. Events with missing transverse momentum and two same-flavor, oppositely charged, low transverse momentum leptons are selected, and are further categorized by the presence of hadronic activity from initial-state radiation or a topology compatible with vector-boson fusion processes. The data are found to be consistent with predictions from the Standard Model. The

Figure 14a Expected  
10.17182/hepdata.91374.v5/t1  
Expected 95% CL exclusion sensitivity for simplified models of direct higgsino production.

Figure 14a Observed  
10.17182/hepdata.91374.v5/t2  
Observed 95% CL exclusion sensitivity for simplified models of direct higgsino production.

Figure 14b Expected  
10.17182/hepdata.91374.v5/t3  
Expected 95% CL exclusion sensitivity for simplified models of direct wino-bino production, assuming  $m(\tilde{\chi}_0^0) \times m(\tilde{\chi}_1^0) < 0$ .

Figure 14b Observed  
10.17182/hepdata.91374.v5/t4  
Observed 95% CL exclusion sensitivity for simplified models of direct wino-bino production.

Version 5 modifications: Fixed analysis labels

Figure 14a Expected <https://www.hepdata.net> JSON

Figure 14a  
Expected 95% CL exclusion sensitivity for simplified models of direct higgsino production.

cmenergies 13000

phrases SUSY Proton-Proton Scattering Exclusion Limits

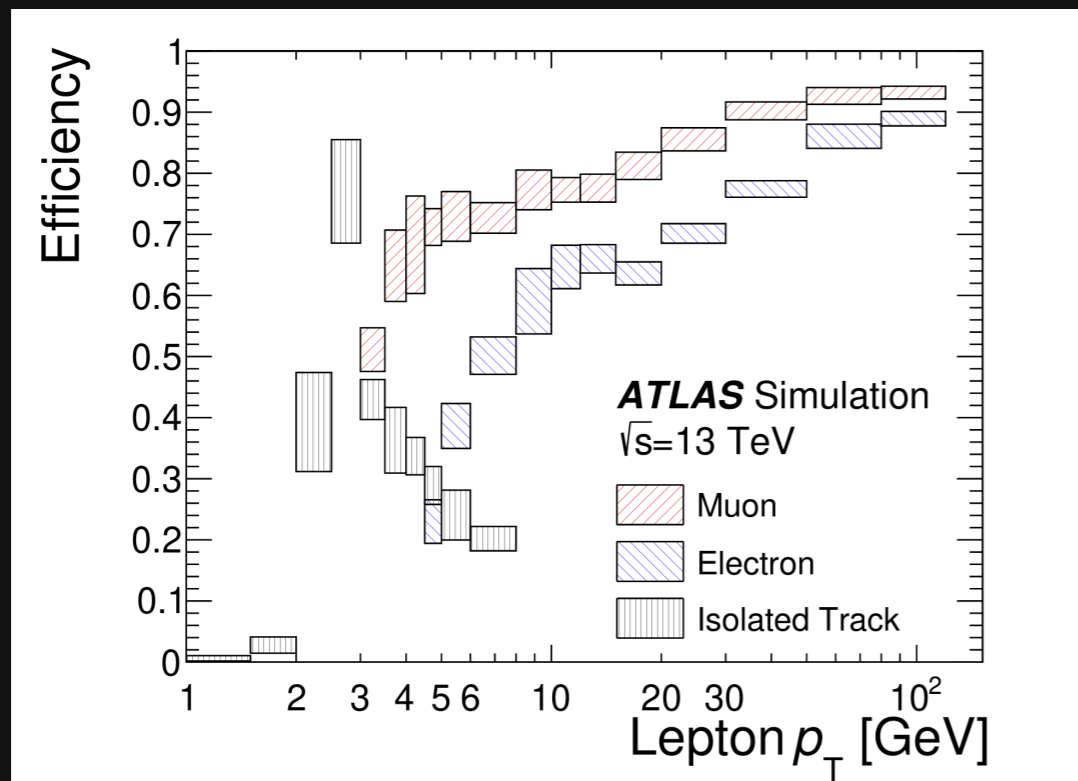
reactions P P -> HIGGSINO HIGGSINO

Showing 50 of 187 values Show All 187 values Visualize

$m(\tilde{\chi}_0^0)$ [GeV]	$\Delta m(\tilde{\chi}_0^0, \tilde{\chi}_1^0)$ [GeV]
81.0	1.6869
82.907	1.712
83.111	1.7147
85.222	1.7434
87.333	1.7738
88.048	1.7844
89.444	1.8055

- HEPData (right) contains:
  - selection acceptances and detector efficiencies for soft leptons (below)
  - HistFactory probability models in a JSON format
  - Exclusion contours
  - Cutflows/Yields

- In addition, SimpleAnalysis contains the generator-level selections (below) for this analysis
  - <https://simpleanalysis.docs.cern.ch/analyses/>
  - Provided as a C++ file, pre-compiled docker images available from ATLAS



```

ANA-SUSY-2018-16.cxx 27.65 KiB
Open in Web IDE Lock Replace Delete

1 #include "SimpleAnalysisFramework/AnalysisClass.h"
2 #include <unordered_set>
3 #include <algorithm>
4
5 //-----
6 //
7 // SUSY EWK Compressed 'Higgsino' Analysis
8 // ANA-SUSY-2018-16
9 //
10 //-----
11
12 DefineAnalysis(EwkCompressed2018)
13
14 void EwkCompressed2018::Init()
15 {
16
17 //Electroweakino exclusive SRs
18 //SR-E-LT, exclusive mll bins (0.5, 1, 1.5, 2, 3, 4, and 5 GeV)
19 addRegions( {"SR_E_LT_eMLLa", "SR_E_LT_eMLLb", "SR_E_LT_eMLLc", "SR_E_LT_eMLLd", "SR_E_LT_eMLLe", "SR_E_LT_eMLLf"}
20 //SR-E-high, exclusive mll bins (1, 2, 3, 5, 10, 20, 30, 40, and 60 GeV)
21 addRegions( {"SR_E_high_eMLLa", "SR_E_high_eMLLb", "SR_E_high_eMLLc", "SR_E_high_eMLLd", "SR_E_high_eMLLe", "SR_E_
22 //SR-E-med, exclusive mll bins (1, 2, 3, 5, 10, 20 and 30 GeV)
23 addRegions( {"SR_E_med_eMLLa", "SR_E_med_eMLLb", "SR_E_med_eMLLc", "SR_E_med_eMLLd", "SR_E_med_eMLLe", "SR_E_med_e
24 //SR-E-low, exclusive mll bins (1, 2, 3, 5, 10, 20, 30, 40, and 60 GeV)
25 addRegions( {"SR_E_low_eMLLa", "SR_E_low_eMLLb", "SR_E_low_eMLLc", "SR_E_low_eMLLd", "SR_E_low_eMLLe", "SR_E_low_e
26
27 //Electroweakino inclusive SRs
28 //SR-E-LT, inclusive mll bins (0.5, 1, 2, 3, 5, 10, 20, 30, 40, and 60 GeV)
29 addRegions( {"SR_E_LT_iMLLa", "SR_E_LT_iMLLb", "SR_E_LT_iMLLc", "SR_E_LT_iMLLd", "SR_E_LT_iMLLe", "SR_E_LT_iMLLf",

```

# So how do we proceed?

## ✓ **Have public data for the analysis, and a -enabled workflow**

- images/containers for MadGraph, Pythia, MadSpin, Delphes, SimpleAnalysis, pyhf, and intermediate output conversions
- ATLAS documented SimpleAnalysis (for truth-level selections + deriving detector acceptances/selection efficiencies)
- pyhf for statistical fitting
- mapyde for chaining all of this together

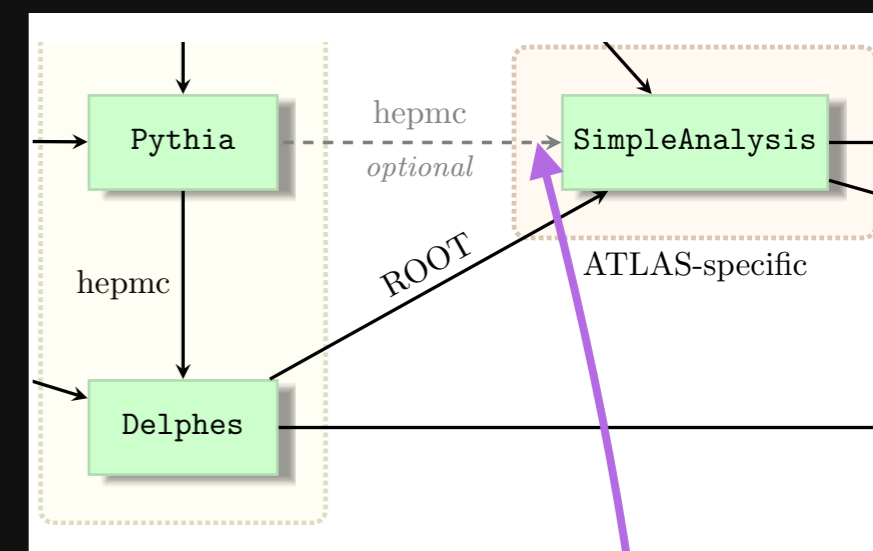
## ✓ **Reproduction of existing results**

- tune differences between our setup and the ATLAS setup (event generation, reconstruction, acceptances and efficiencies, cross-section)

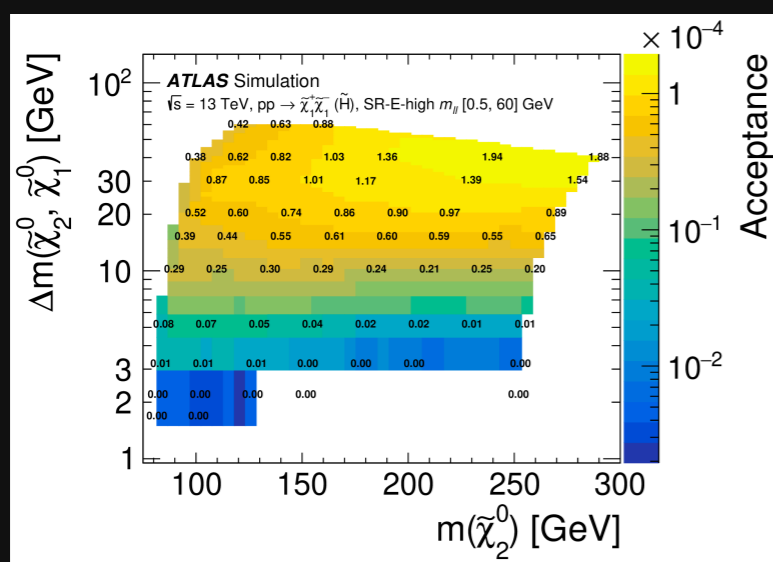
## ✓ **Reinterpretation of (new) alternate theories**

- for each model considered (sleptons and Higgsinos), identify a new model we want to explore current experimental sensitivity

# Acceptance Comparisons



- ATLAS has a different setup for producing their signal models: MadGraph with  $\leq$  **2-jets + MadSpin**
  - Detector Reconstruction/Simulation: **ATLAS Internal**
- Our paper uses: MadGraph with  $\leq$  **1-jet + Pythia**
  - note: mapyde supports MadSpin (we opted not to use it)*
  - Detector Reconstruction/Simulation: **Delphes** (with default ATLAS card)
- Compare the acceptances from our workflow to acceptances published by ATLAS — at most 10% difference (in the compressed-mass region)



$$A \times \varepsilon \times BR \times \sigma \times \mathcal{L}_{int}$$

detector acceptance  $\times$  selection efficiency  $\times$  branching ratio  $\times$  cross section  $\times$  integrated luminosity



# Tuning Cross-Section

Cross-sections for all but the VBF signal scenarios are calculated with **RESUMMINO 2.0.1** at **NLO+NLL** precision [63–70]. The VBF cross-sections are computed at LO precision with **MG5\_aMC@NLO 2.6.2**. The evaluation of the cross-sections and corresponding uncertainty are taken from an envelope of cross-section predictions using different PDF sets, and varied factorization and renormalization scales. This procedure is described in Ref. [71], and is the same procedure as used in the previous search [45].

- In this analysis, ATLAS published NLO cross-sections calculated with **Resummino**
- Our cross-sections are calculated at LO ( $\leq 1$  jet) with **MadGraph**
- Derive a **k-factor** for each signal point that **we apply** in order to adjust for quantum (loop) corrections

$$k \equiv \frac{\sigma_{\text{NLO}}}{\sigma_{\text{LO}, 0 \text{ jet}}}$$

← ATLAS provides  $\sigma_{\text{NLO}}$   
 ← Special mapyde run with 0-jet at LO  $\sigma_{\text{LO}, 0 \text{ jet}}$

$$\sigma_{\text{eff}} = k \times \sigma_{\text{LO}, 1 \text{ jet}}$$

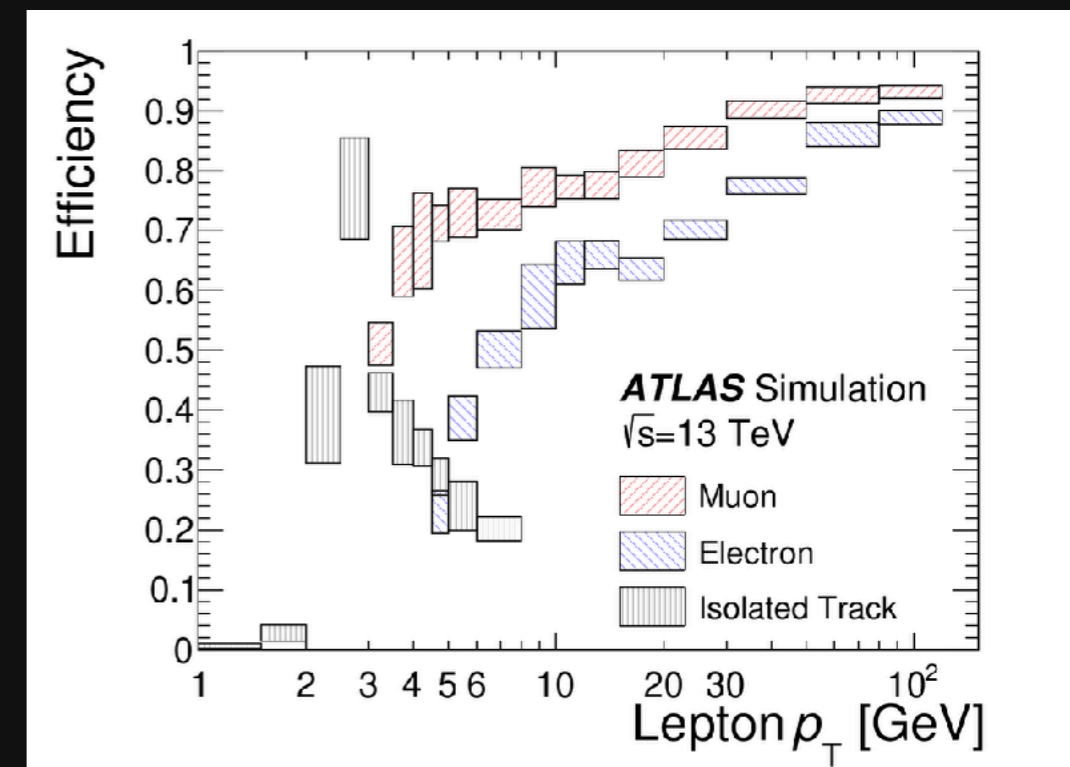
**mass-independent  $k = 1.18$**

# Tuning Reconstruction

- Then, tune default ATLAS card by incorporating the (soft) lepton reconstruction efficiencies

```

set EfficiencyFormula {
    (pt < 4.5) * (0.00) +
    (abs(eta) <= 2.5) * (pt < 5.0) * (pt >= 4.5) * (0.30) +
    (abs(eta) <= 2.5) * (pt < 6.0) * (pt >= 5.0) * (0.45) +
    (abs(eta) <= 2.5) * (pt < 8.0) * (pt >= 6.0) * (0.52) +
    (abs(eta) <= 2.5) * (pt < 10) * (pt >= 8.0) * (0.65) +
    (abs(eta) <= 2.5) * (pt < 20) * (pt >= 10) * (0.68) +
    (abs(eta) <= 2.5) * (pt < 30) * (pt >= 20) * (0.70) +
    (abs(eta) <= 2.5) * (pt < 50) * (pt >= 30) * (0.75) +
    (abs(eta) <= 2.5) * (pt >= 50) * (0.87) +
    (abs(eta) > 2.5) * (0.00)}
    
```



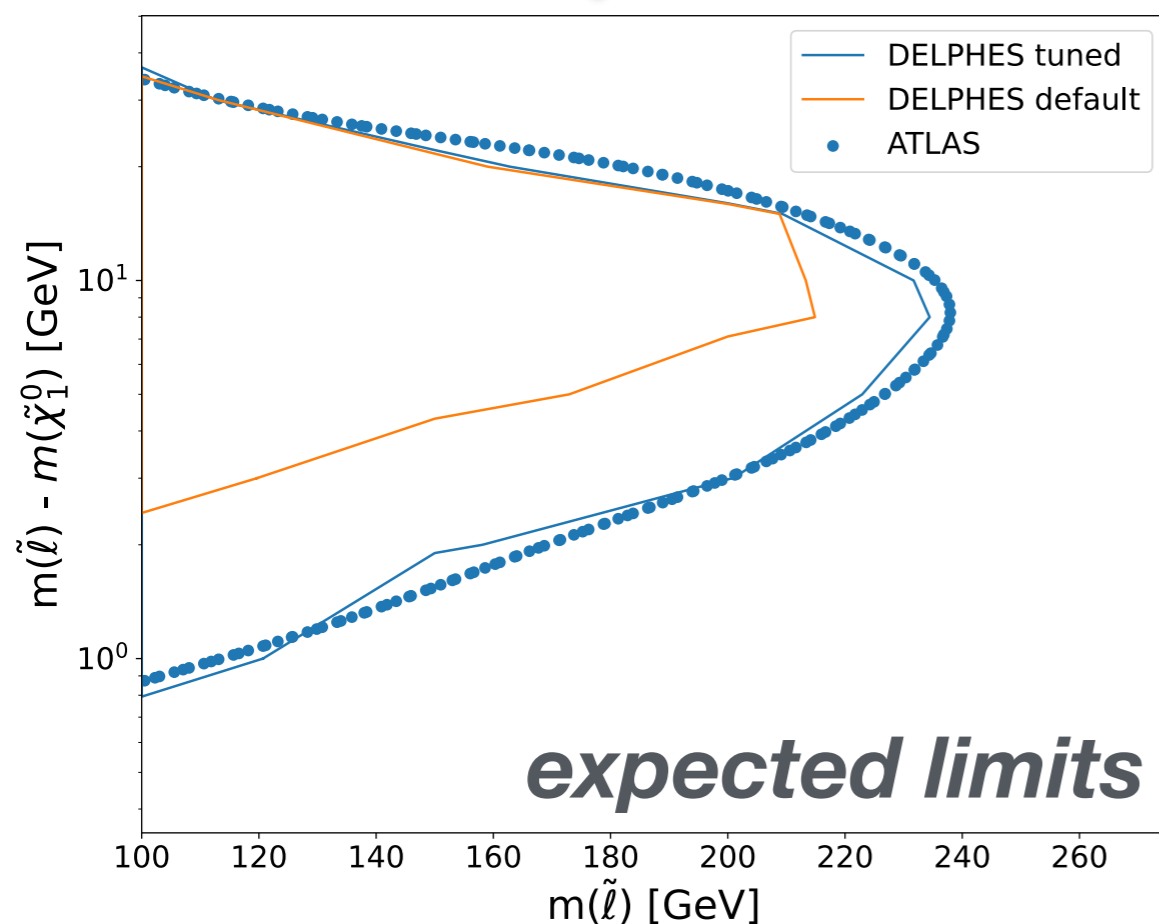
reconstructing (very) soft leptons

- This means updating the TCL (config) that Delphes uses include comment on upper bound

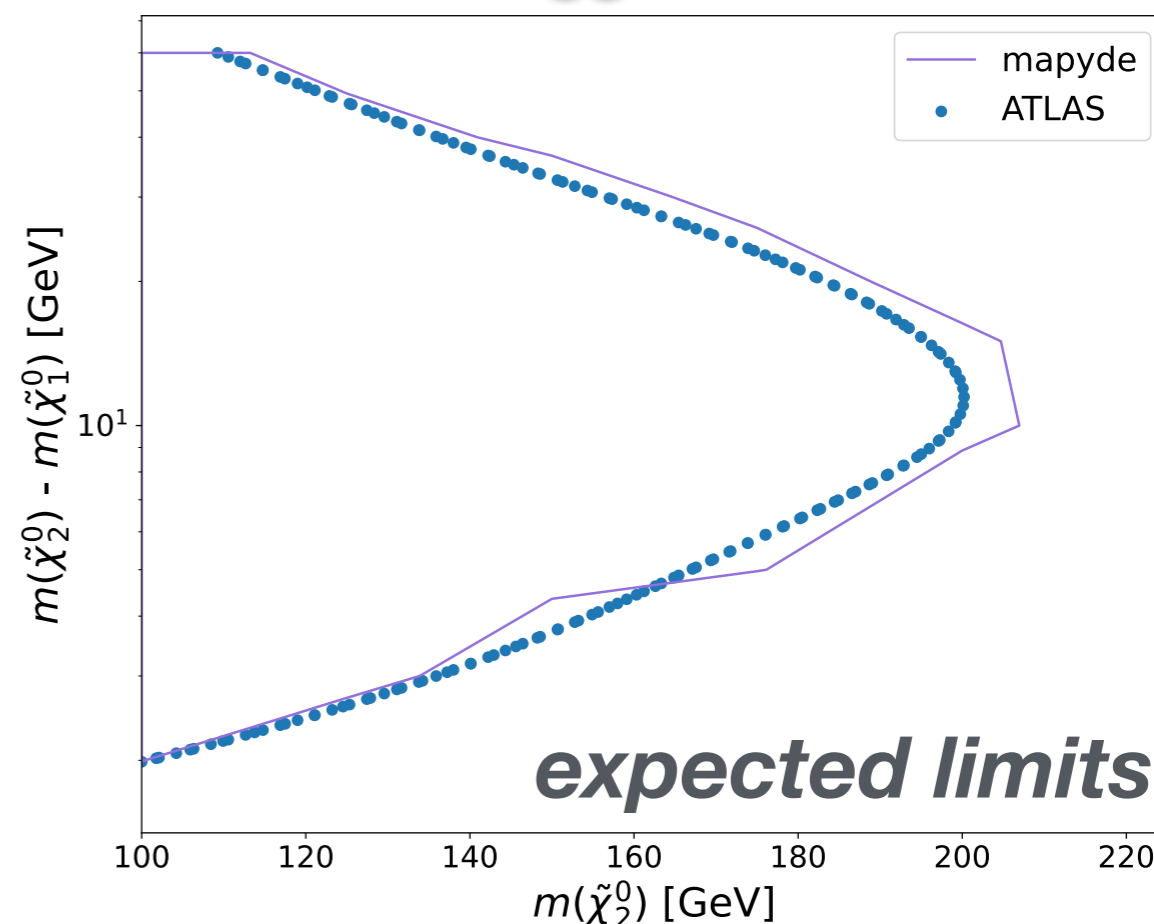


# Reproduction

sleptons



Higgsinos



- Sleptons: lepton reconstruction efficiencies are an essential piece of the workflow

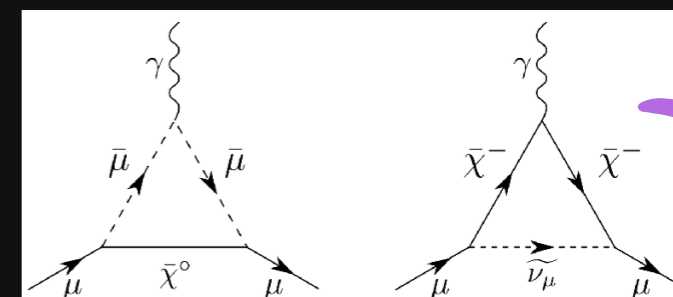
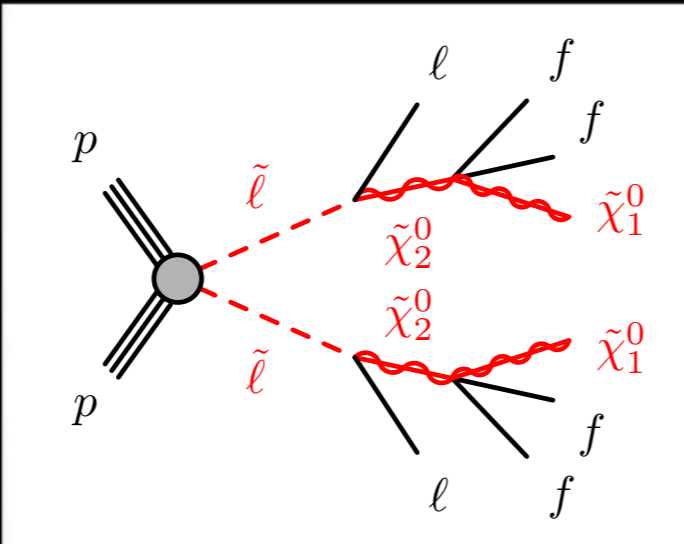
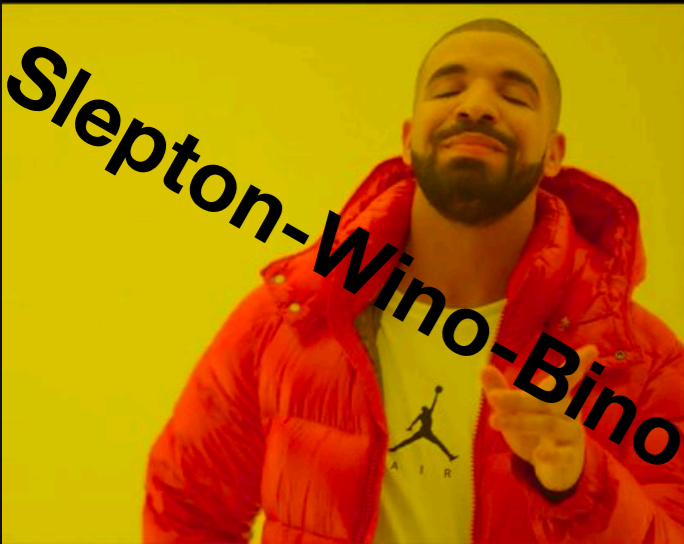
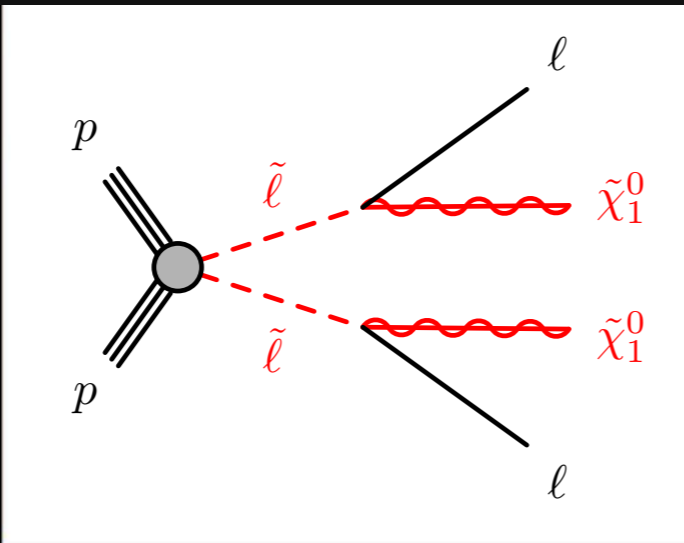
- Higgsinos: also includes lepton reconstruction efficiencies (just not overlaid)

✓ Our workflow... works.

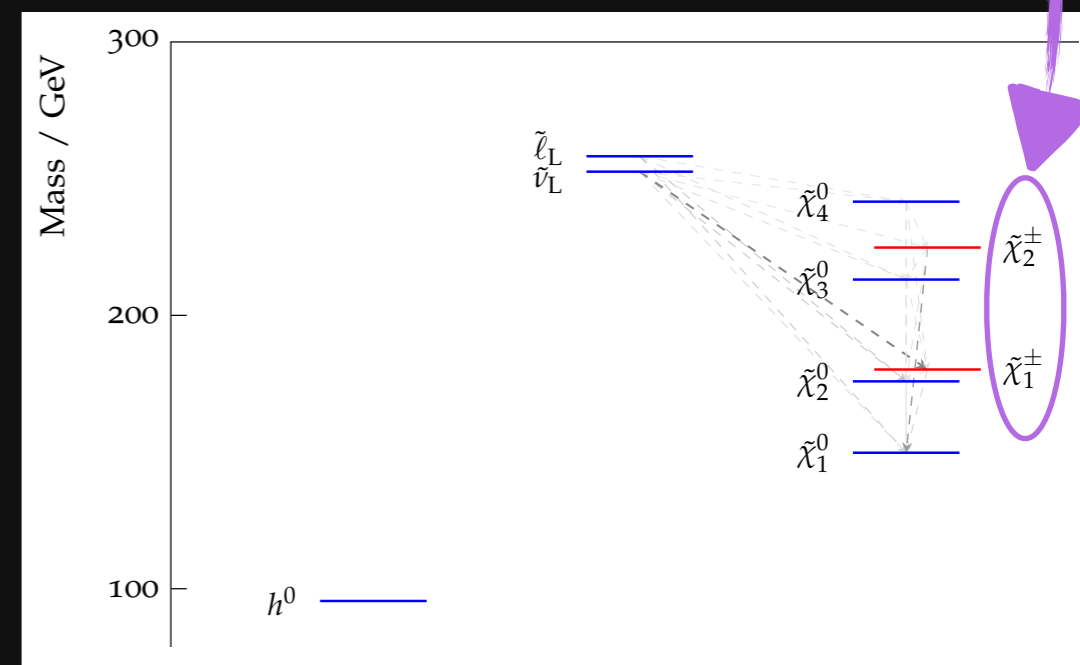


mapyde

# Reinterpretation (sleptons)



$\tilde{\mu}^\pm / \tilde{\chi}^0$  or  $\tilde{\chi}^\pm / \tilde{\nu}_\mu$



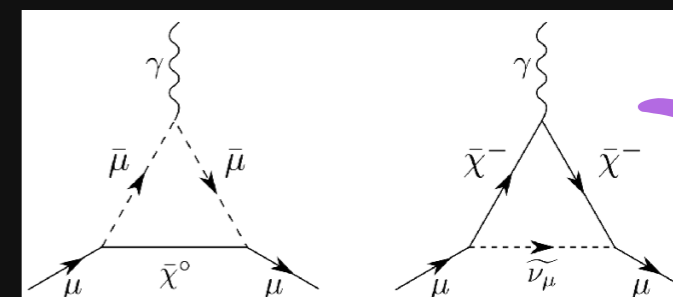
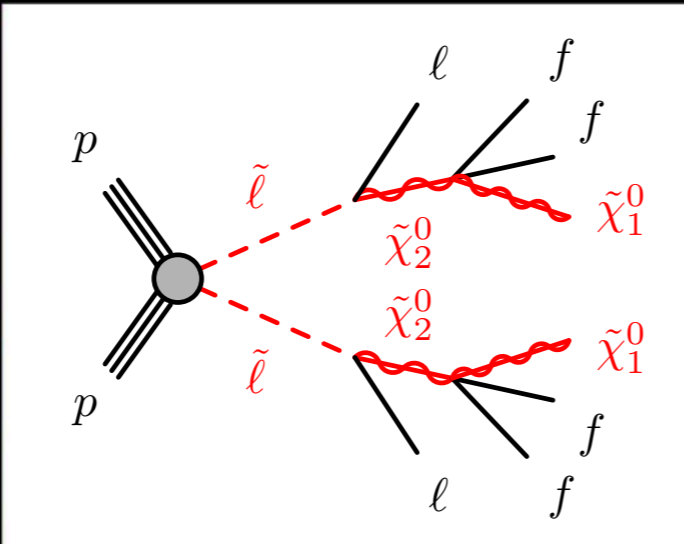
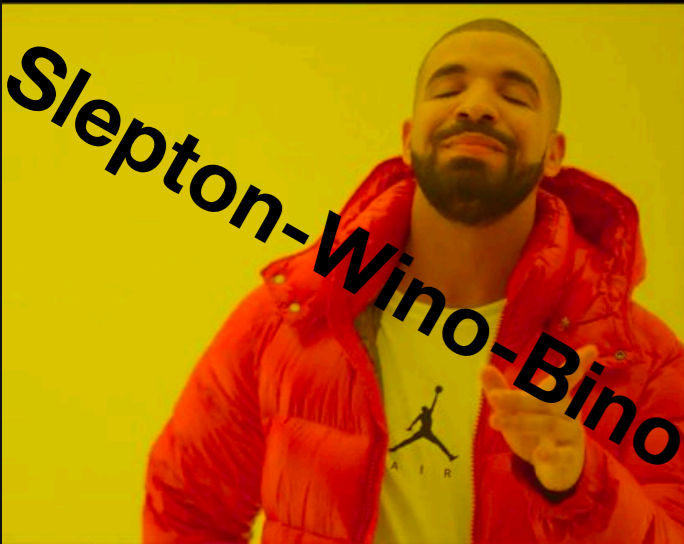
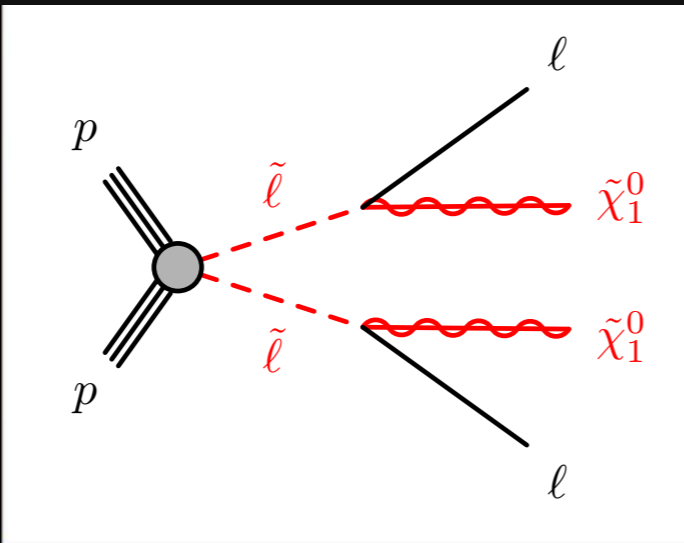
ATLAS did a “simplified model” search — is it too simple?

Including wino expands parameter-space, perhaps contributions to g-2?

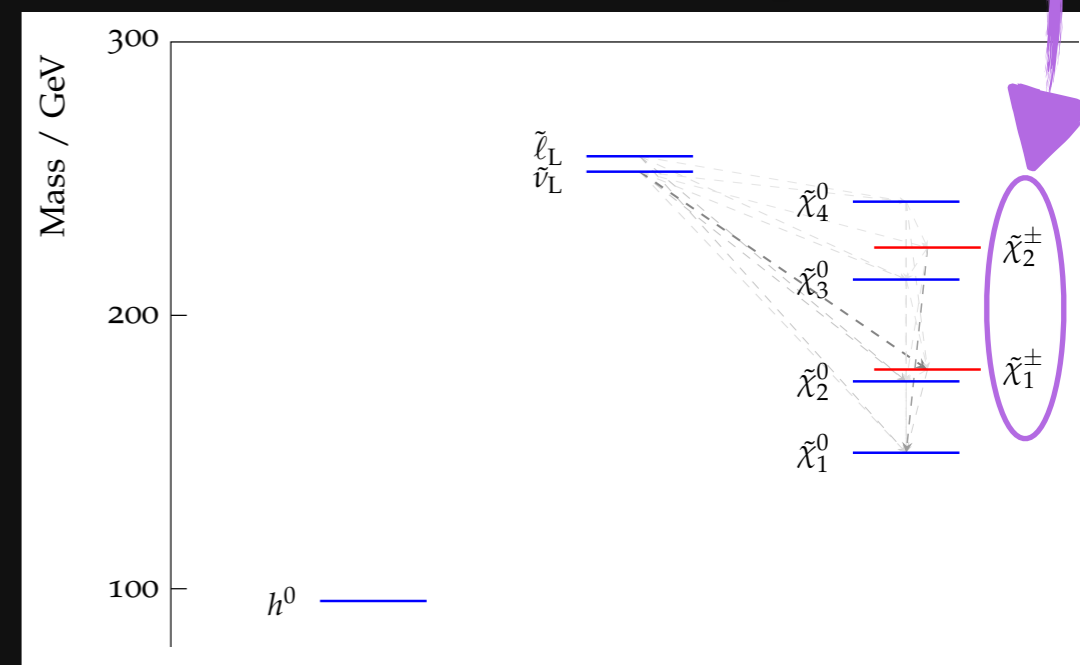


mapyde

# Reinterpretation (sleptons)



$\tilde{\mu}^\pm / \tilde{\chi}^0$  or  $\tilde{\chi}^\pm / \tilde{\nu}_\mu$

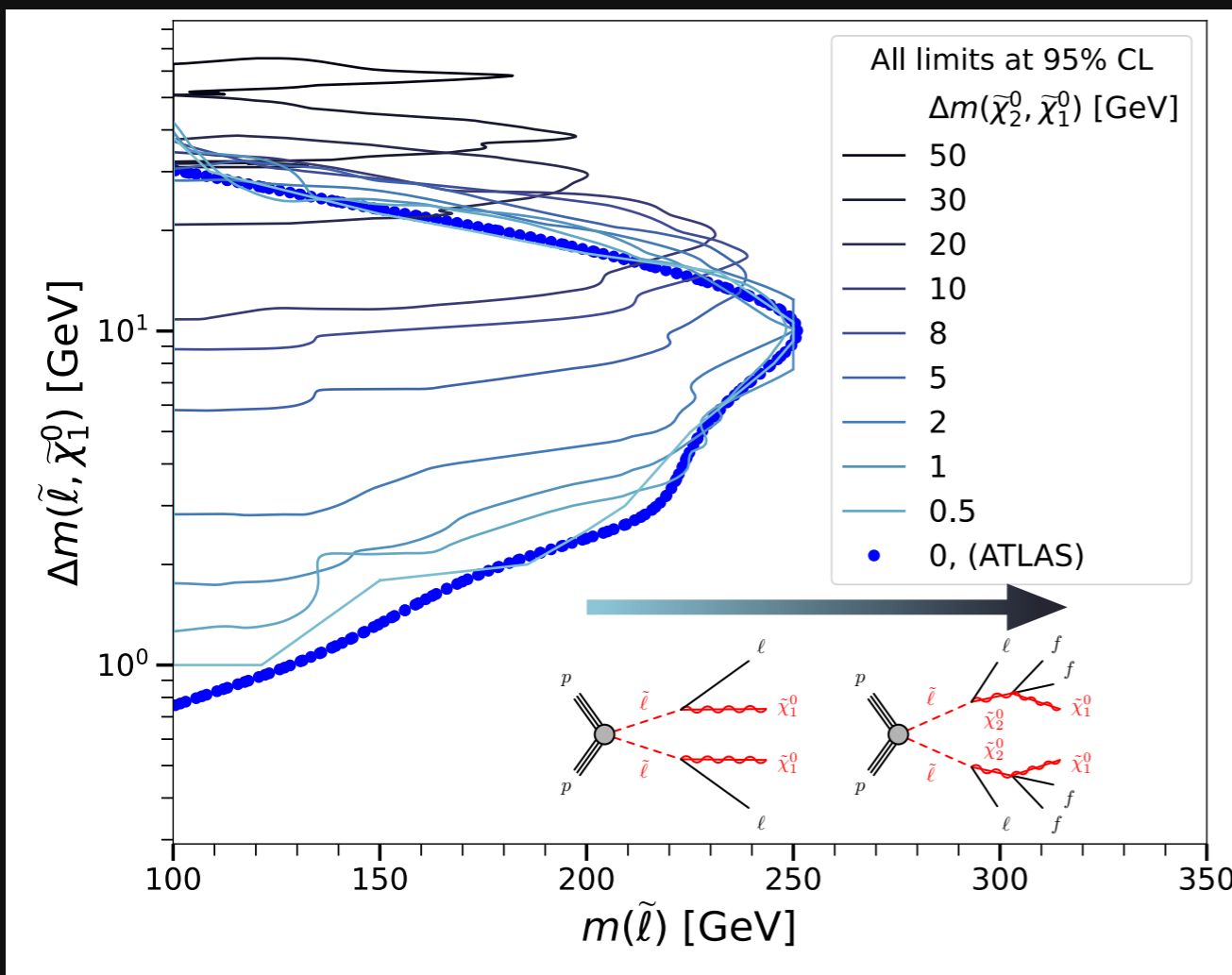


- Including wino expands parameter-space, perhaps contributions to g-2?

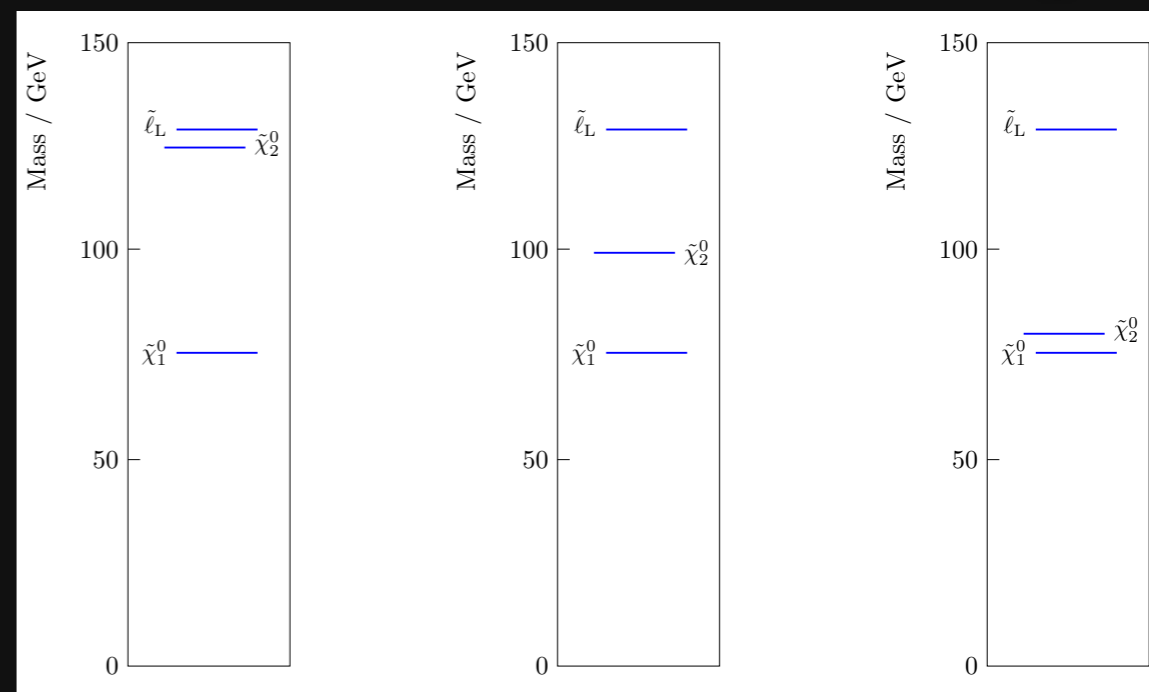


# Reinterpretation (sleptons)

## observed limits



- lepton  $p_T$  determined by gap of:
  - slepton-bino:  $\tilde{\ell} - \tilde{\chi}_1^0$
  - slepton-wino-bino:  $\tilde{\ell} - \tilde{\chi}_2^0$

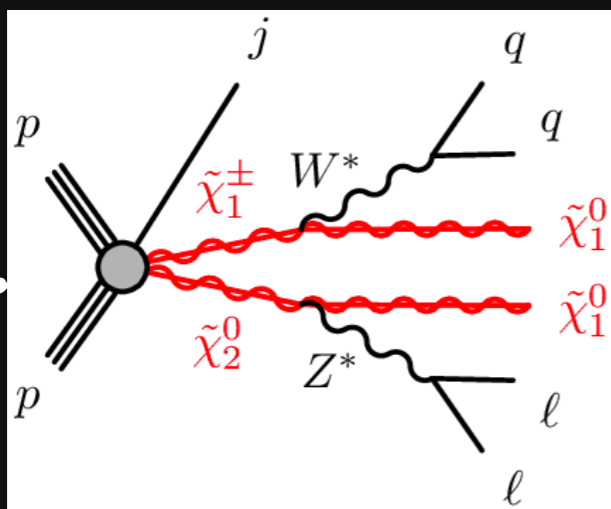


- each contour is bounded:
  - below**: weakened sensitivity at smaller  $\tilde{\ell} - \tilde{\chi}_2^0$  due to softer SM leptons/lower lepton efficiencies
  - above**: reduced  $\sigma$  due to larger slepton masses, and selection efficiency (optimized for softer)

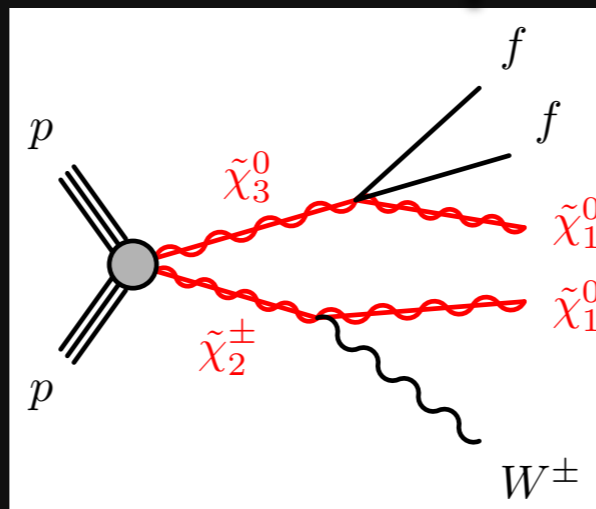
- at  $\tilde{\chi}_2^0 - \tilde{\chi}_1^0$  splittings that are:
  - small**: resembles slepton-bino ATLAS results (degeneracy of neutralinos)
  - large**: stronger constraints at large  $\tilde{\ell} - \tilde{\chi}_1^0$

# Reinterpretation (WBH)

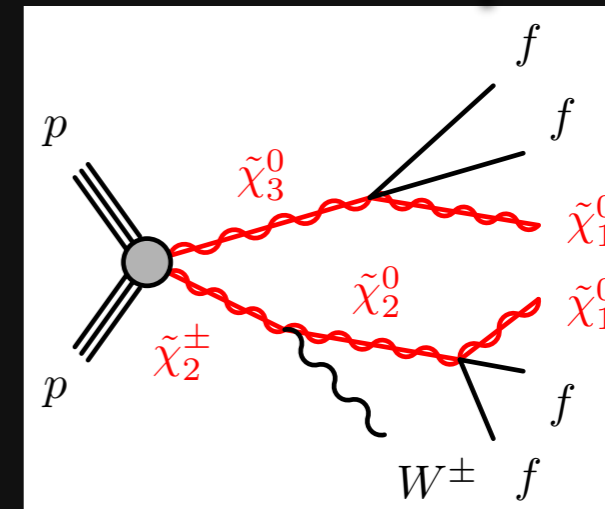
“higgsino”  
ATLAS



“wino-bino-higgsino” (WBH)  
one-step



two-step

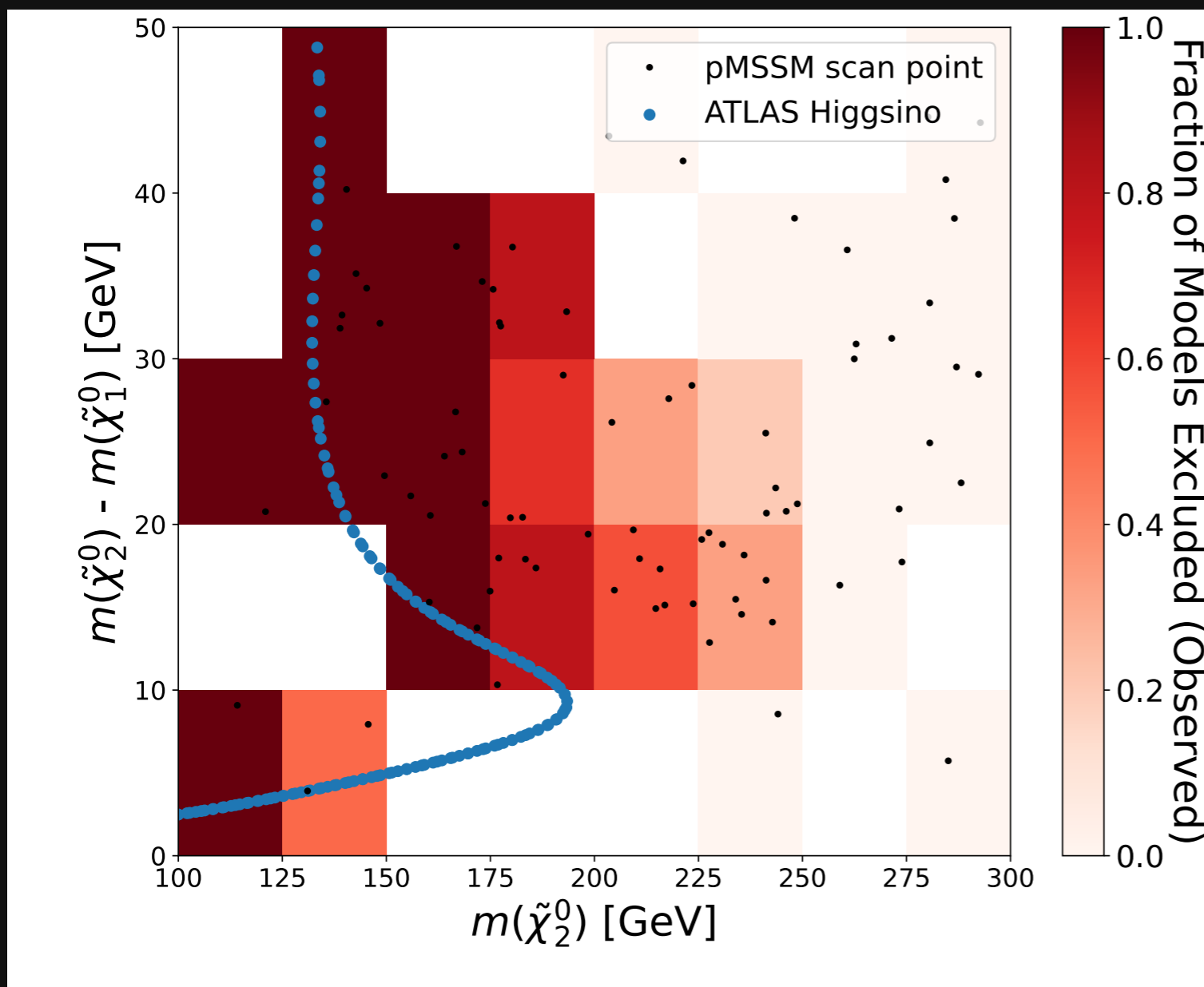


ATLAS looked Higgsinos  $(\tilde{\chi}_1^0, \tilde{\chi}_1^\pm, \tilde{\chi}_2^0)$  where the neutralinos were compressed...

**What if** the wino-bino terms  $(M_1, M_2)$  were “close by”  $\mu$  so that we have an additional neutralino and chargino  $(\tilde{\chi}_2^\pm, \tilde{\chi}_3^0)$ ?

**! Let's run a pMSSM scan and see?**

# Reinterpretation (WBH)



- ✦ Scan 500k points, 81 models survived
  - ✦  $m(\tilde{\chi}_1^0) > 100$  GeV
  - ✦  $m(\tilde{\chi}_3^0) < 300$  GeV
  - ✦  $\Delta m(\tilde{\chi}_3^0, \tilde{\chi}_1^0) < 50$  GeV
  - ✦ SPheno: valid
  - ✦ FeynHiggs: Higgs mass
  - ✦ Micromegas: Dark matter
  - ✦ SuperIso: Flavor physics
  - ✦ GM2Calc: g-2
  
- ✦ Compared to a pure-Higgsino model, the presence of an additional neutralino  $\tilde{\chi}_3^0$  with similar decays to  $\tilde{\chi}_2^0$  might be helping

*observed limits*

**! Stronger constraints at larger mass splittings!**



# The tech stack



🐍 “pip”-installable (also exists on conda-forge)

🚢 docker/singularity/apptainer are the runtime environments (must exist on the host machine or node)

📄 configuration (apart from individual cards used by various tools) is in TOML

- ✦ modern configuration that’s human-readable

# TOML?

top-level folder to store outputs

output folder name  
(uses *Jinja2* templating)

```
[base]
path = "/data/users/mhance/SUSY"
output = "SUSY_SleptonWinoBino_isrslep_{{madgraph['masses']['MSLEP']}}_{{madgraph['masses']['MN2']}}_{{madgraph['masses']['MN1']}}"
template = "{{PWD}}/templates/sleptons.toml"
process_path = "{{PWD}}/cards/process/"
param_path = "{{PWD}}/cards/param/"

[madgraph]
params="SleptonWinoBino"
nevents=50000

[madgraph.proc]
name = "isrslep"

[madgraph.masses]
MSLEP = xxx
MC1 = xxx
MN2 = xxx
MN1 = xxx
MSNU = xxx

[pythia]
mpi = "on"
```

template allows for "inheritance"  
(makes it easy to build "grids")

param card (SLHA)

process card

values to substitute in param card

 mapyde run user.toml

# TOML?

top-level folder to store outputs

output folder name (uses Jinja2 templating)

```

[base]
path = "/data/users/mhance/SUSY"
output = "SUSY_SleptonWinoBino_isrslep_{{madgraph['masses']['MSLEP']}}_{{madgraph['masses']['MN2']}}_{{madgraph['masses']['MN1']}}.tar.gz"
template = "{{PWD}}/template"
process_path = "{{PWD}}/process"
param_path = "{{PWD}}/parameters"

[base]
name = "isrslep"
card = """import model MSSM_SLHA2
define chsleptons = e1- e1+ er- er+ mu1- mu1+ mu2- mu2+
generate          p p > chsleptons chsleptons j
output -f"""

[base]
name = "isrslep"

[base]
MSLEP = xxx
MC1 = xxx
MN2 = xxx
MN1 = xxx
MSNU = xxx

[base]
mpi = "on"

```

[madgraph.proc] can also inline process card!

```

name = "isrslep"
card = """import model MSSM_SLHA2
define chsleptons = e1- e1+ er- er+ mu1- mu1+ mu2- mu2+
generate          p p > chsleptons chsleptons j
output -f"""

```

ce")  
s")

process card

values to substitute in param card

 mapyde run user.toml

# Custom usage: e.g. rivet

```
import mapyde
from mapyde.utils import build_config, load_config
from mapyde.container import Container

filename = "user.toml"
user = load_config(filename)
config = build_config(user) ← parse config, handle inheritance,
                             and template substitutions

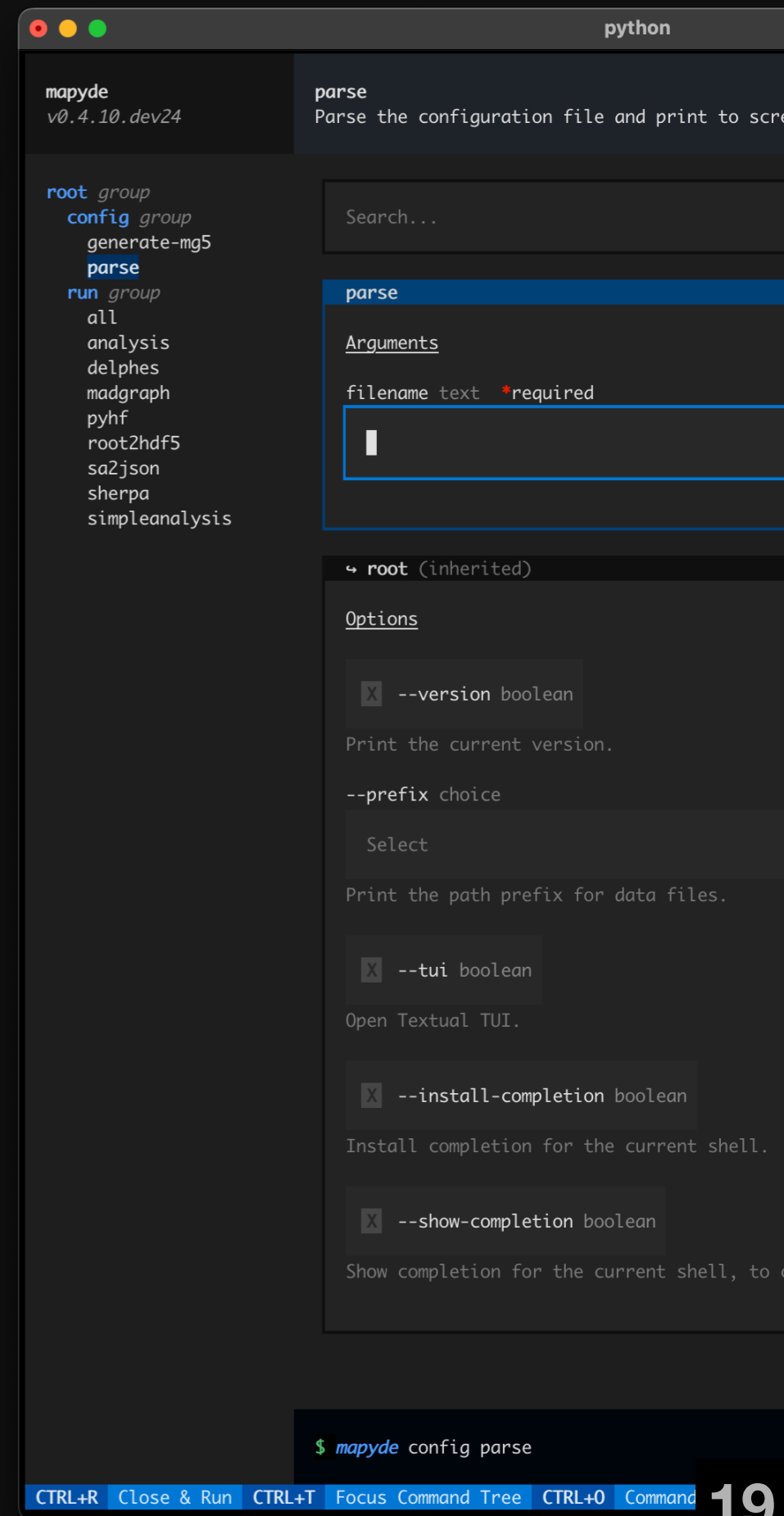
image = "docker.io/hepstore/rivet:latest"
command = bytes(
    f"""mkfifo fifo.hepmc && generator -o fifo.hepmc && rivet
-a {config['rivet']['name']} fifo.hepmc""",
    "utf-8",
)

with Container( ← container wrapper to run the
                command in the image
    image=image,
    name=f"{config['base']['output']}__rivet",
    engine=config["base"].get("engine", "docker"),
    mounts=mounts(config),
    stdout=sys.stdout,
    cwd="/data",
    output_path=utils.output_path(config),
    logs_path=config["base"]["logs"],
) as container:
    stdout, stderr = container.call(command)
```

- ✦ out of the box, mapyde comes with enough utilities to roll your own workflow
- ✦ maybe you want to use something that's not currently provided

# Other technical bits

- **mapyde utilities**: <https://scipp-atlas.github.io/mapyde/latest/reference/mapyde/utis/>
  - merging configs, rendering Jinja2 template strings, handling inheritance
- **mapyde prefix**: <https://scipp-atlas.github.io/mapyde/latest/reference/mapyde/prefix/>
  - “global” configuration paths for data, cards, scripts, templates
  - can be used in python within a context manager (to change paths temporarily)
- **mapyde runner**: <https://scipp-atlas.github.io/mapyde/latest/reference/mapyde/runner/>
  - hard-coded support for MadGraph, Pythia, Delphes, Sherpa, SimpleAnalysis, SimpleAnalysis->JSON, pyhf, ROOT->HDF5, and custom analysis scripts
  - can provide support for other tools upon request in GitHub issues
- **mapyde tui** (textual user interface) from command line



```
python
mapyde v0.4.10.dev24
parse Parse the configuration file and print to screen

root group
  config group
    generate-mg5
    parse
  run group
    all
    analysis
    delphes
    madgraph
    pyhf
    root2hdf5
    sa2json
    sherpa
    simpleanalysis

Search...



parse
Arguments
filename text *required

root (inherited)
Options
[X] --version boolean
Print the current version.
--prefix choice
Select
Print the path prefix for data files.
[X] --tui boolean
Open Textual TUI.
[X] --install-completion boolean
Install completion for the current shell.
[X] --show-completion boolean
Show completion for the current shell, to...

$ mapyde config parse
```

CTRL+R Close & Run CTRL+T Focus Command Tree CTRL+O Command

# Summary

- A **reinterpretation** of compressed electroweak SUSY was performed using public ATLAS data and mapyde — **could guide Run 3 searches**
- Relies on many pieces in place from both experimental and theory communities, and from tech industry — **successful coordination**
- Low barrier to entry means mapyde can be used as a **pedagogic tool** for teaching new students particle physics through a hands-on approach
- **paper:** [arXiv:2306.11055](https://arxiv.org/abs/2306.11055) (submitted to SciPost Phys)
  - **repo:**  [scipp-atlas/mapyde-paper](https://github.com/scipp-atlas/mapyde-paper)
  - **data:**  [scipp-atlas/mapyde-paper-data](https://github.com/scipp-atlas/mapyde-paper-data)

docs **online**

License **Apache 2.0**

DOI **10.5281/zenodo.8040910**

pypi package **0.5.0**

conda-forge **v0.5.0**

Backup

# pMSSM Scan

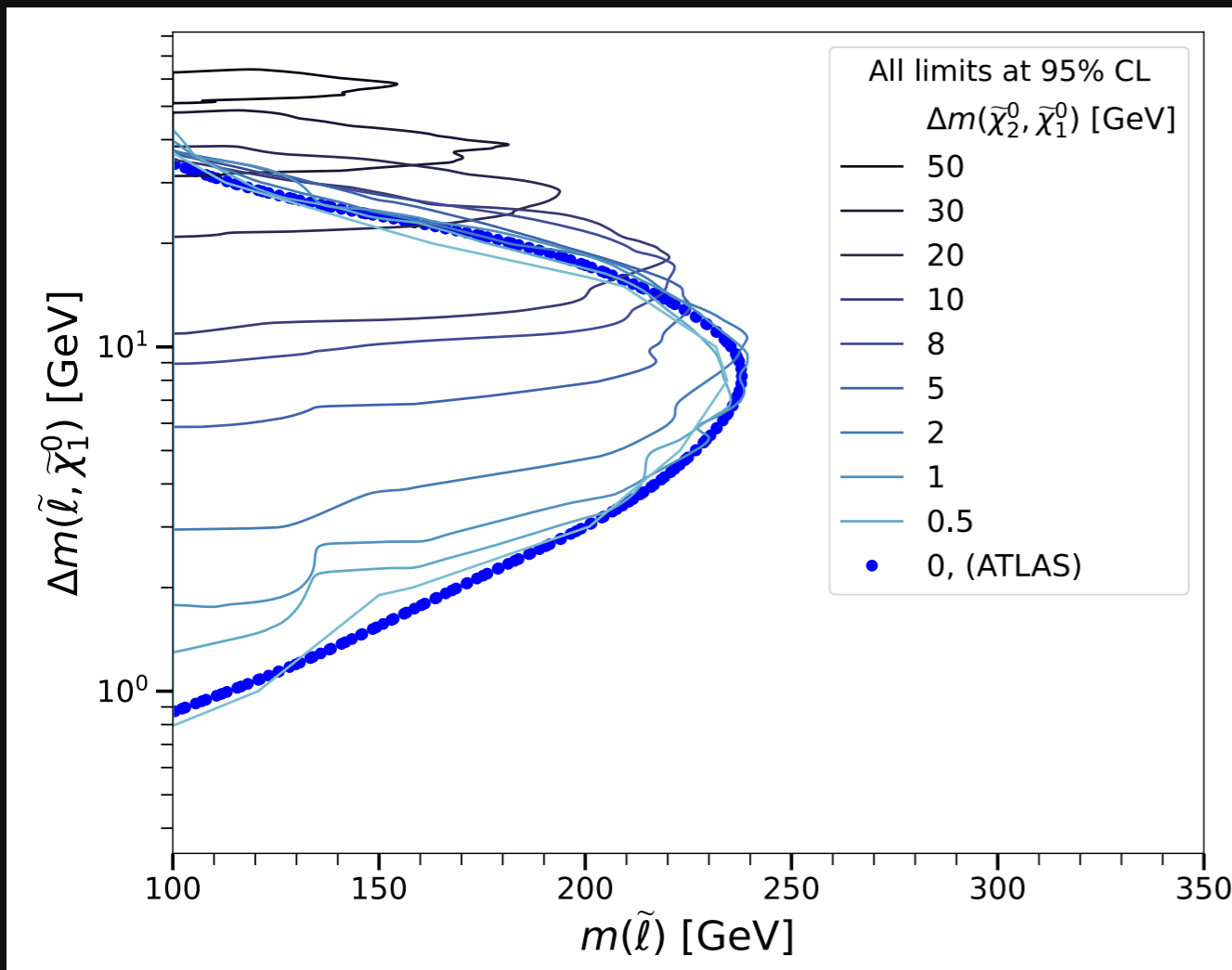
- EASYSCAN\_HEP (v1.0.0): pMSSM scanning and program control [48, 49]
- SPHENO (v4.0.4): spectrum generator [50, 51]
- FEYNHIGGS (v2.16.0): Higgs mass calculation [53–60]
- MICROMEAS (v5.2.1): Dark Matter calculations (e.g., relic density) [61]
- SUPERISO (v4.0): Flavor Physics observables [62]
- GM2CALC (v2.0.0):  $g - 2$  calculation [63, 64]

```
[scan]
Scan method:      random
#                ID      Prior  Min    MAX
Input parametes: tanb,   Flat,   1,     60
                  M_1,   Flat,  -500,  500
                  M_2,   Flat,  -500,  500
                  M_3,   Flat,  2000, 2000
                  AT,    Flat,  2000, 2000
                  Ab,    Flat,  2000, 2000
                  Atau,  Flat,  2000, 2000
                  MU,    Flat,  -500,  500
                  mA,    Flat,  2000, 2000
                  meL,   Flat,  2000, 2000
                  mtauL, Flat,  2000, 2000
                  meR,   Flat,  2000, 2000
                  mtauR, Flat,  2000, 2000
                  mqL1,  Flat,  2000, 2000
                  mqL3,  Flat,  2000, 2000
                  muR,   Flat,  2000, 2000
                  mtR,   Flat,  2000, 2000
                  mdR,   Flat,  2000, 2000
                  mbR,   Flat,  2000, 2000
```

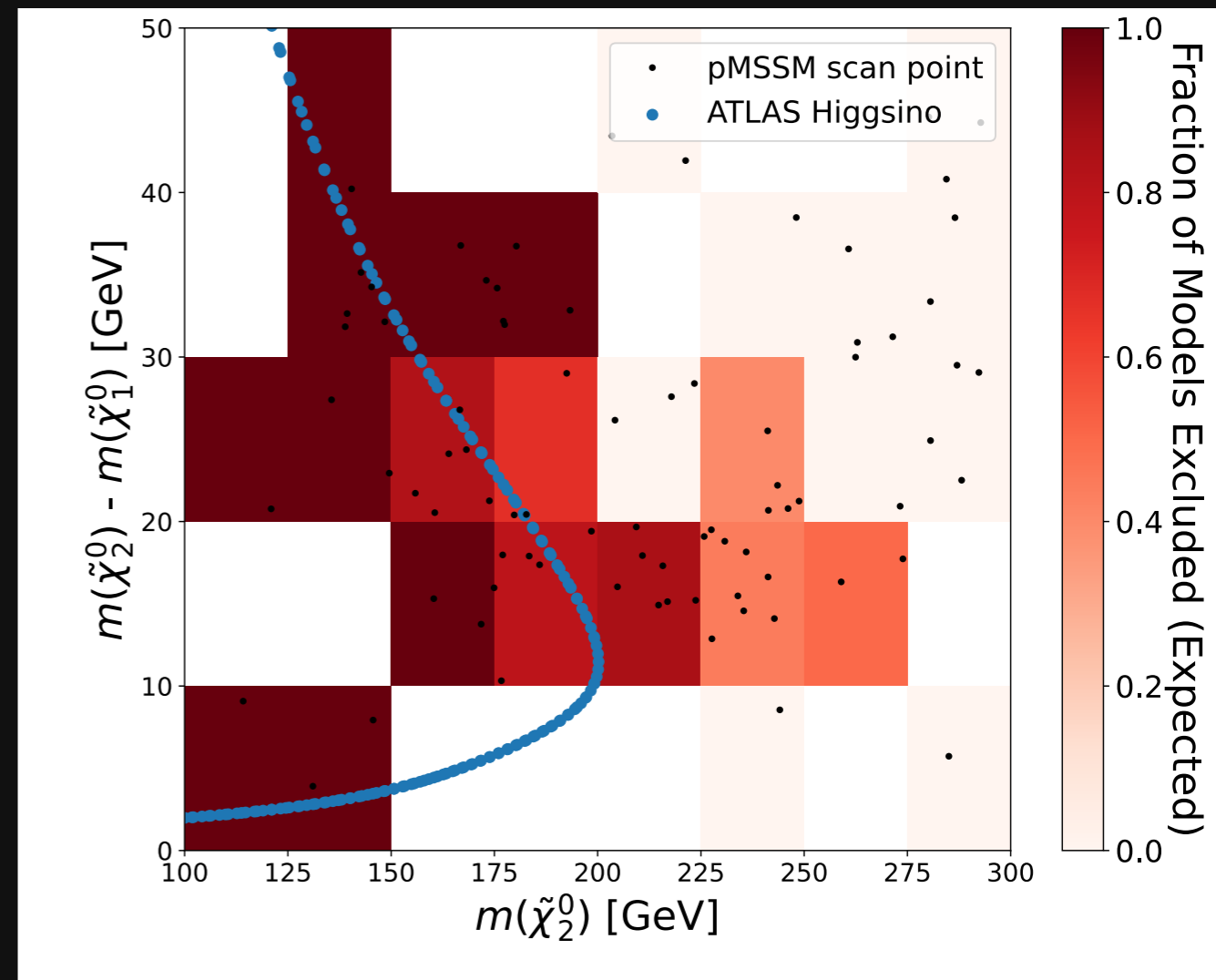
Listing 1: A portion of the easyscan.ini configuration defining the random sampling for the electroweakinos scan.



# Expected



sleptons



wino-bino-higgsino