

Progress towards a more sensitive CWoLa hunt with the ATLAS detector

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1 Introduction

2 Strategy

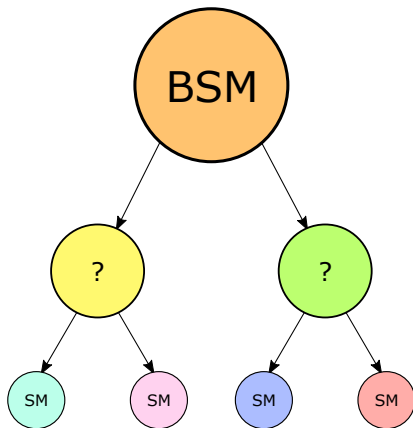
- CWoLa and its problems
- SALAD
- Strategic confusion

3 Projected sensitivity

4 Conclusion

Model-agnostic searches

- Wide range of model-specific searches
- Even wider range of possible signal models
- Model-agnostic searches can dramatically expand search 'width'



Data used

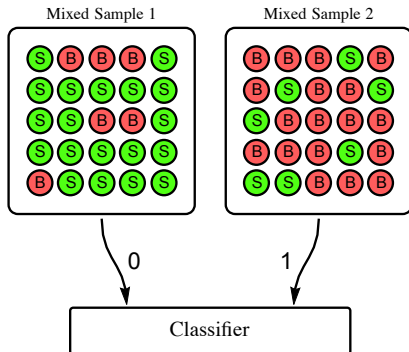
- High p_T dijets from ATLAS Run 2 data
 - 139 fb^{-1}
- Blinding:
 - Considering events with $|\Delta\eta| > 1.2$
 - Only looking at 10% of the above
- Simulation is Pythia

CWoLa overview



CWoLa

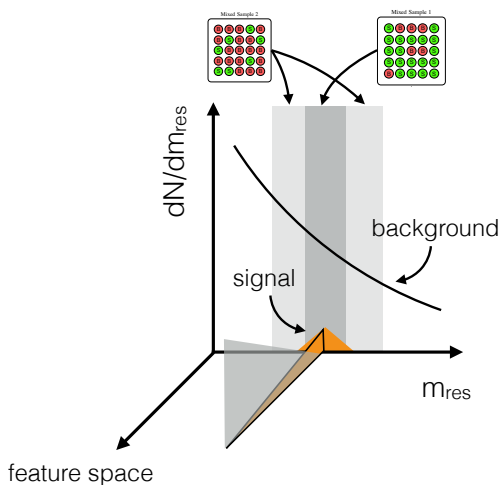
*Classification
Without Labels*



- Classifier learns likelihood ratio
- See arXiv: [1708.02949](https://arxiv.org/abs/1708.02949), [1902.02634](https://arxiv.org/abs/1902.02634)

Standard CWoLa hunt

- Previously performed with ATLAS Run 2 data
 - Looking at dijet events
- **Resonant feature:** dijet invariant mass
- **Classification feature:** jet masses
- See arXiv: [2005.02983](https://arxiv.org/abs/2005.02983)

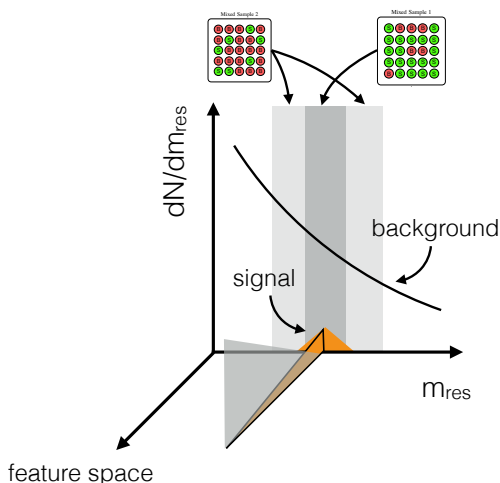


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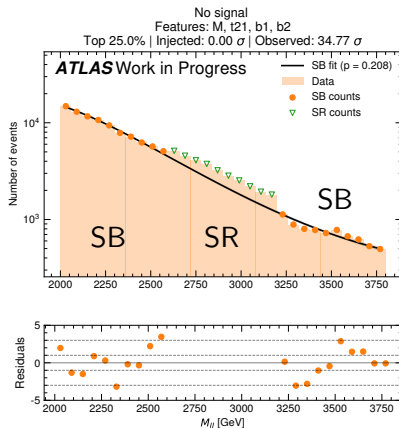
Our Goal

Achieve higher sensitivity with more classification features

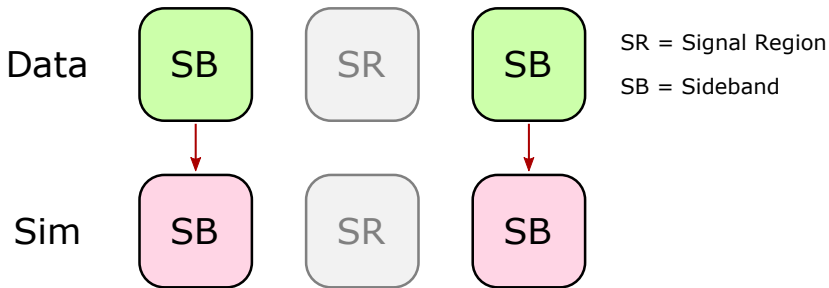


Problems with CWoLa

- Signal region (SR) and sideband (SB) have different M_{JJ}
- More classification features → CWoLa infers M_{JJ}
- CWoLa tags entire signal region as signal
- **Result:** large false positive and low sensitivity to real signal
 - Previously avoided with statistical decorrelation

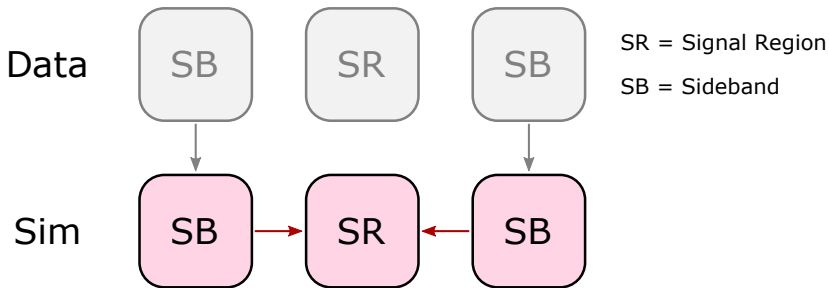


Step 1: learn imperfections in simulation



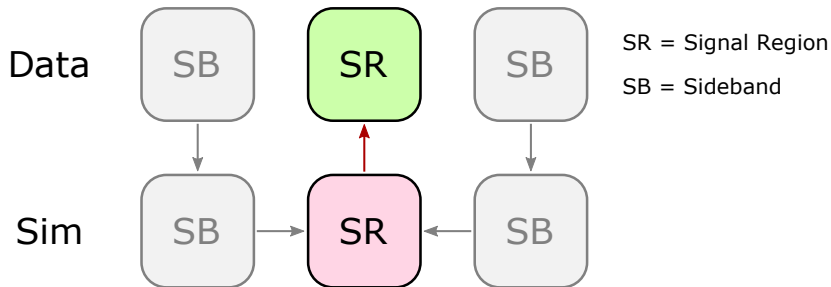
- **SALAD: Simulation Assisted Likelihood-free Anomaly Detection**
- See arXiv: [2001.05001](https://arxiv.org/abs/2001.05001)

Step 2: reweight SR in simulation



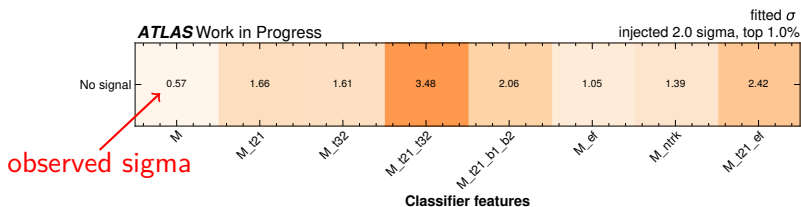
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Step 3: learn simulation vs data in SR



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- See arXiv: [2001.05001](https://arxiv.org/abs/2001.05001)

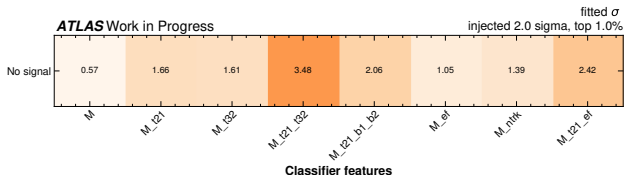
Results without signal



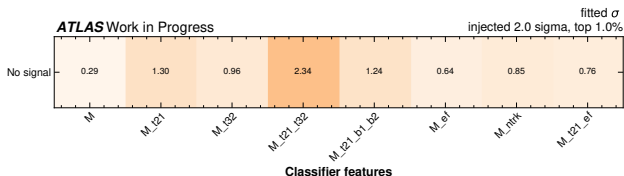
- Colors and values are fitted significance
 - Fit in sideband according to $p_1(1 - m)^{p_2} m^{-p_3}$
 - Interpolate to signal region
- Features used:
 - M : jet mass
 - τ_{21} : N -subjettiness ratio τ_2/τ_1
 - τ_{32} : N -subjettiness τ_3/τ_2
 - b_1, b_2 : b tagging on first and second subjets
 - EF: fraction of jet energy lost to EM calorimeter
 - N tracks: number of jet tracks

Strategic confusion

- Standard SALAD without signal:

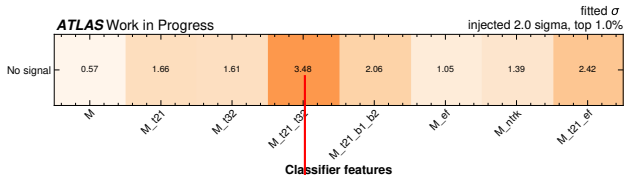


- Injecting some sideband events into signal region can reduce false positive significance
- With 25% of sideband in SR:

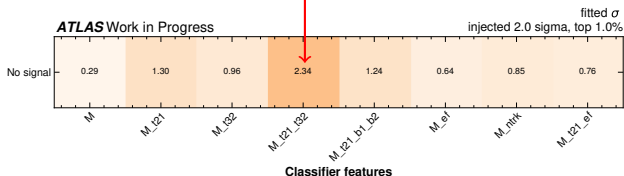


Strategic confusion

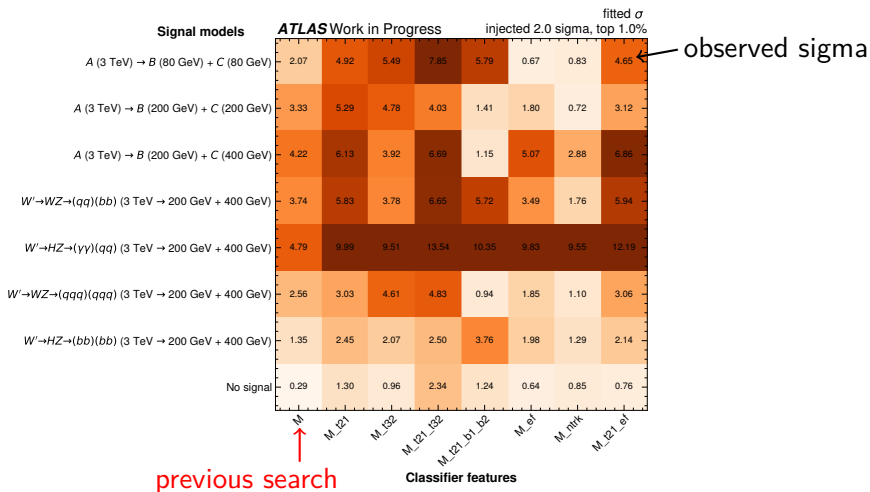
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Sensitivity to 2σ signal



- Search strategy can take 2σ excess to $>5\sigma$ result

Conclusion

- Making progress towards a more sensitive model-agnostic search in ATLAS Run 2 data
 - More classifier features → wider search
 - Simulation assistance (SALAD) → more robust search
 - **Next steps:** need to correct for classification bias
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- **Acknowledgements:** I am very grateful to Benjamin Nachman and Luc Le Pottier for their support, guidance, and collaboration