

# Non-resonant Anomaly Detection

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with background extrapolation

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# Current landscape of Anomaly Detection

## Requirements for complete Anomaly Detection



## Many proposals and all data results



Examples: SALAD, CATHODE, FETA, CURTAINS, ...

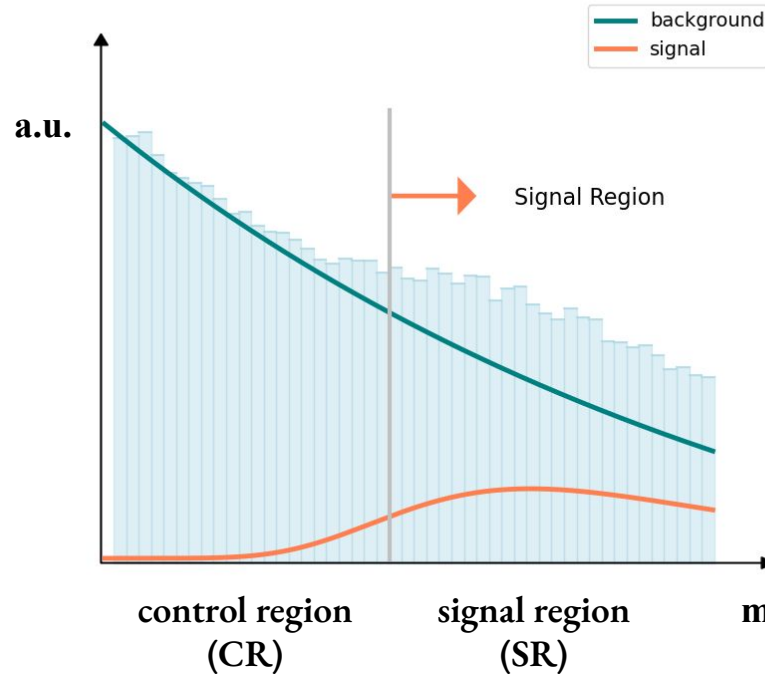
See [Radha's talk](#) for [the interplay of resonant AD methods](#).

# This proposal

Background **extrapolation**  
from one side



**non-resonant** signal

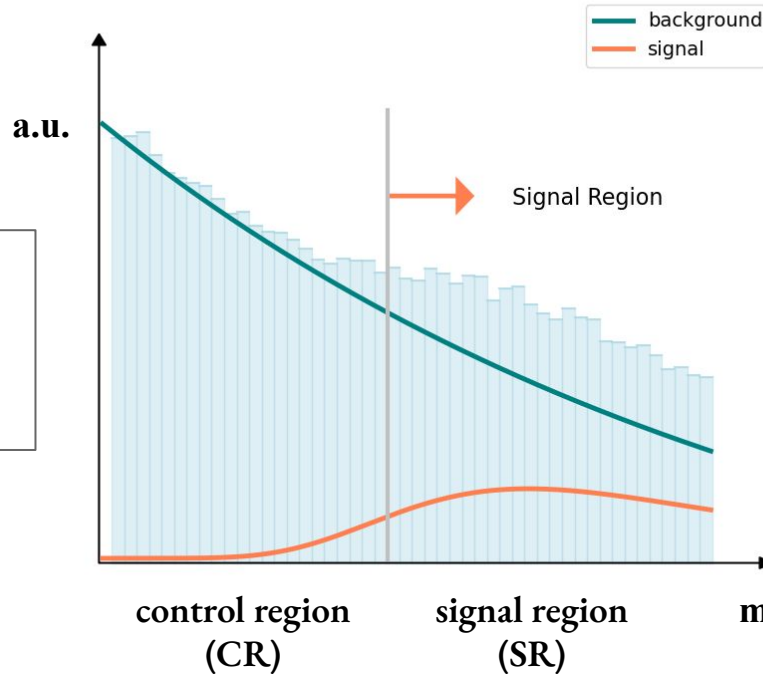


# This proposal

Background **extrapolation**  
from one side



**non-resonant** signal



See [Gerrit's talk](#) on [tail-based AD](#).

Though using interpolation, it also used some non-resonance features.

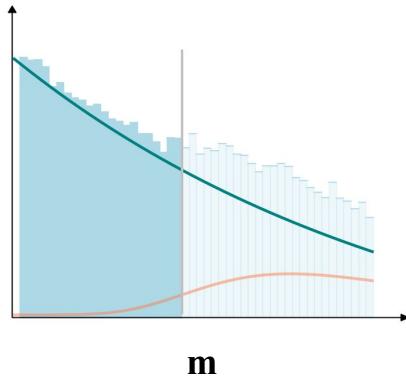
# The idea

→ Why don't we just extrapolate with existing methods?

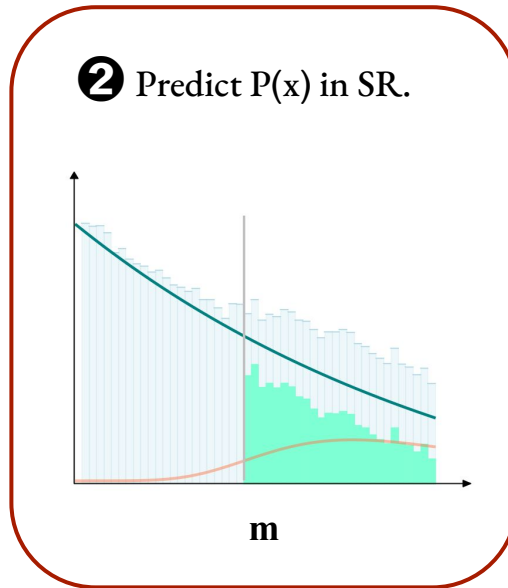
In principle we could use only MC-to-data reweighting (SALAD).

However, density estimation with generative models are **less robust without a sideband**.

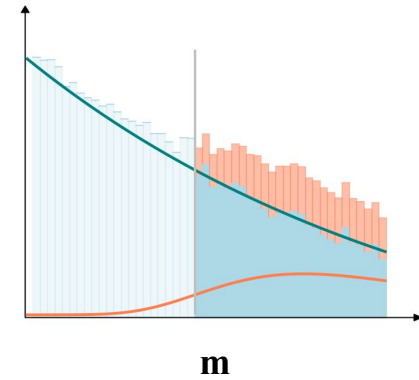
❶ Learn  $P(x|m)$  in CR.



❷ Predict  $P(x)$  in SR.



❸ Classification in SR



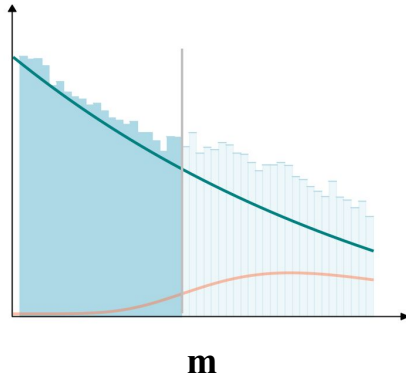
# The idea

→ Hybridize generated models with SALAD reweighting!

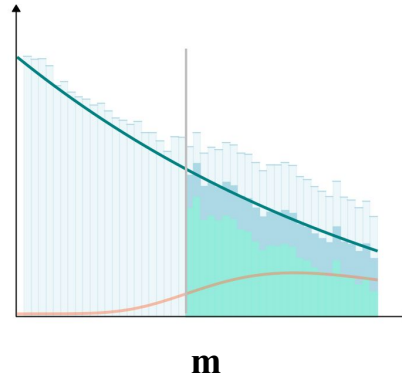
Reweight conditional variables from MC to data in SR.

$$\begin{aligned} P_{\text{data}}(x, m) &= P_{\text{data}}(x|m)P_{\text{data}}(m) \\ &= P_{\text{data}}(x|m)P_{\text{MC}}(m) \frac{P_{\text{data}}(m)}{P_{\text{MC}}(m)} \\ &= P_{\text{data}}(x|m)P_{\text{MC}}(m) w(m) \end{aligned}$$

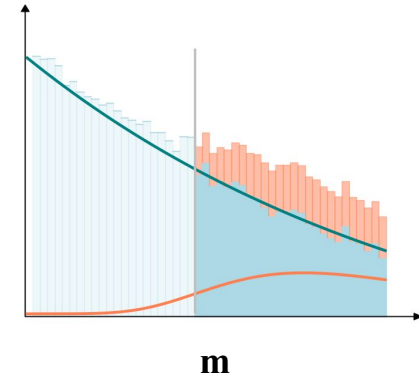
❶ Learn  $P(x|m)$  in CR.



❷ Predict  $P(x)$  in SR with reweighting.

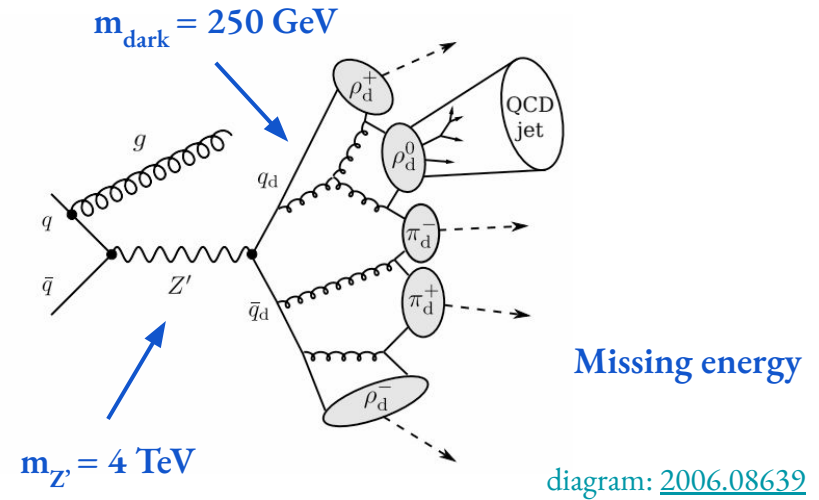
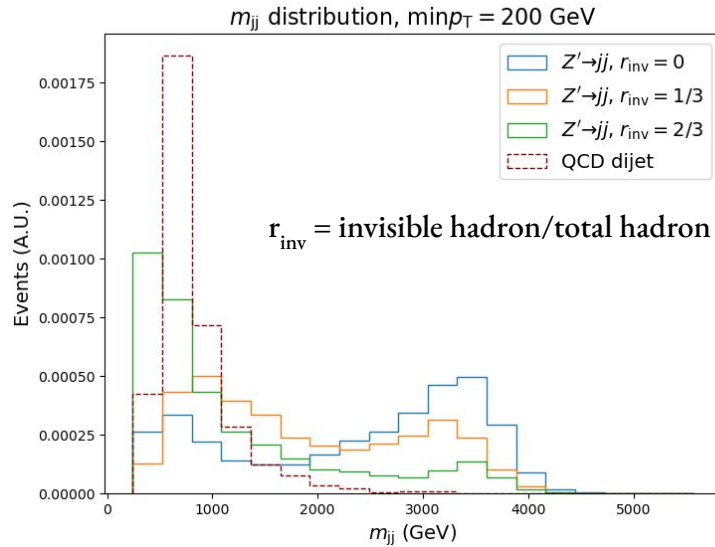


❸ Classification in SR



# Non-resonant signals

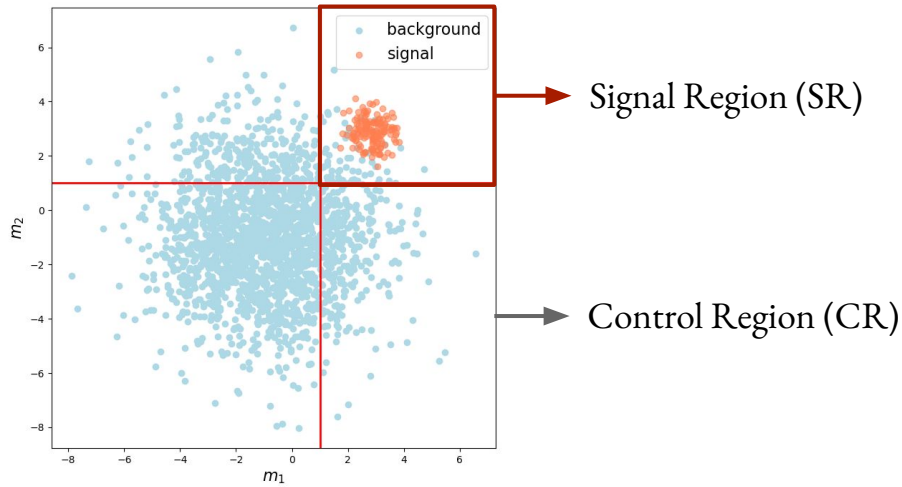
Case 1: large missing energy in the final states of a resonant production. Example here: semi-visible jets from  $Z'$ .



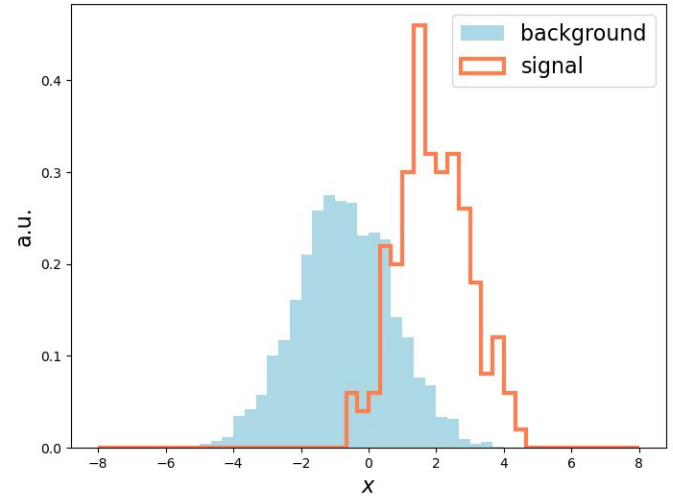
Case 2: off-shell effects from heavy particles. Example: modifications to SMEFT coefficients, but not explored here.

# First, let's take a look at a toy example

Context variables  $m_1$  and  $m_2$ , used to define SR & CR.



A feature variable  $x$ .



Test SALAD-only, CATHODE+SALAD, and FETA+SALAD.



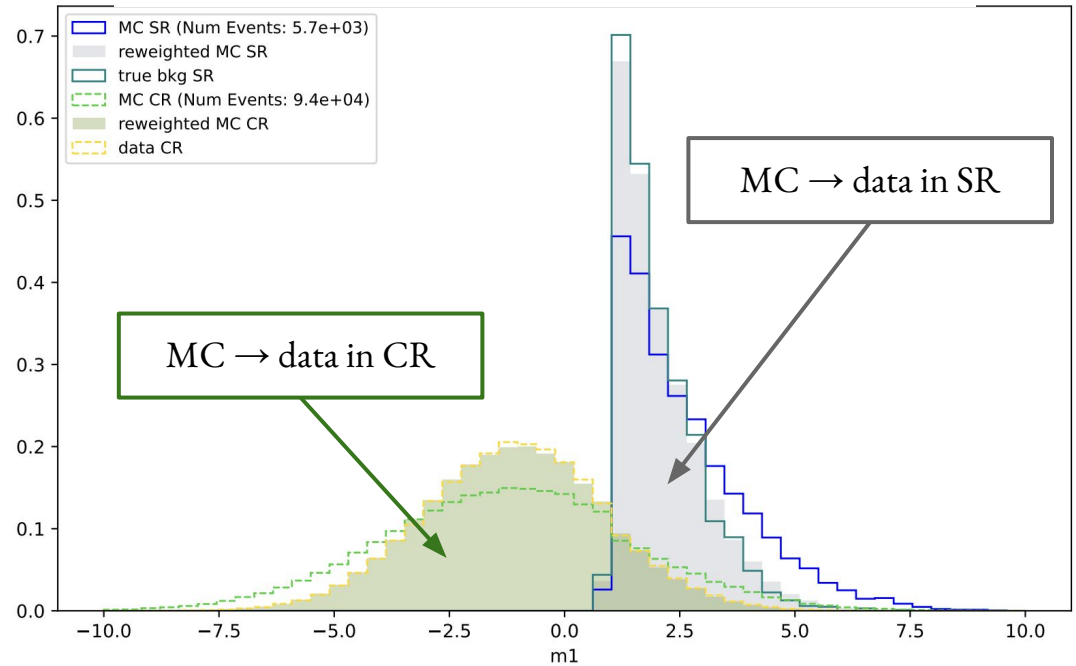
# Anomaly detection methods

- Reweighting (SALAD)

Assumption:

$$w(m) = \frac{P_{data}(m \in CR)}{P_{MC}(m \in CR)}$$
$$= \frac{P_{data}(m \in SR)}{P_{MC}(m \in SR)}$$

Reweighted MC vs data for the toy background

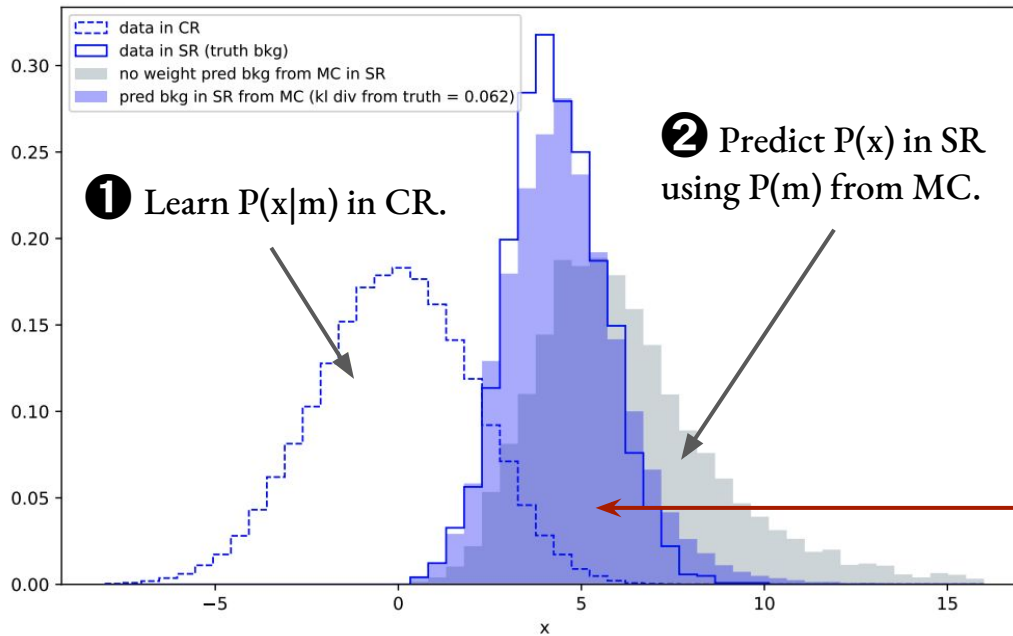


\* This plot is used for illustration purpose only, not the final result.

# Anomaly detection methods

- Density estimation (CATHODE & FETA + SALAD)

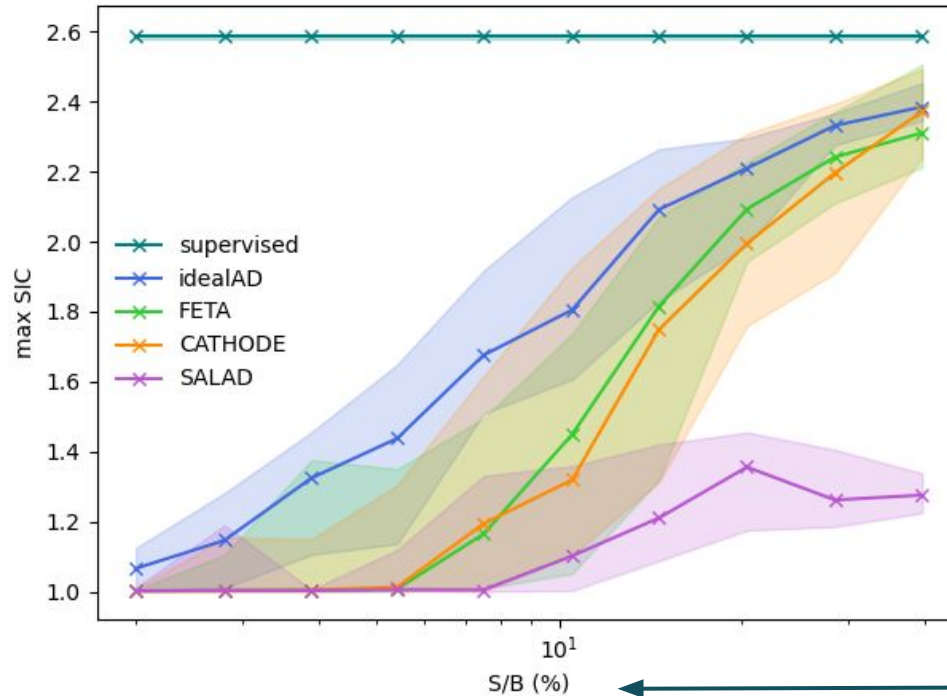
True vs predicted toy background



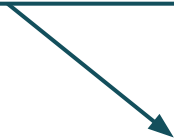
\* This plot is used for illustration purpose only, not the final result.

# Results from the toy example

## AD performance over increasing signal injections



Maximum Significance Improvement

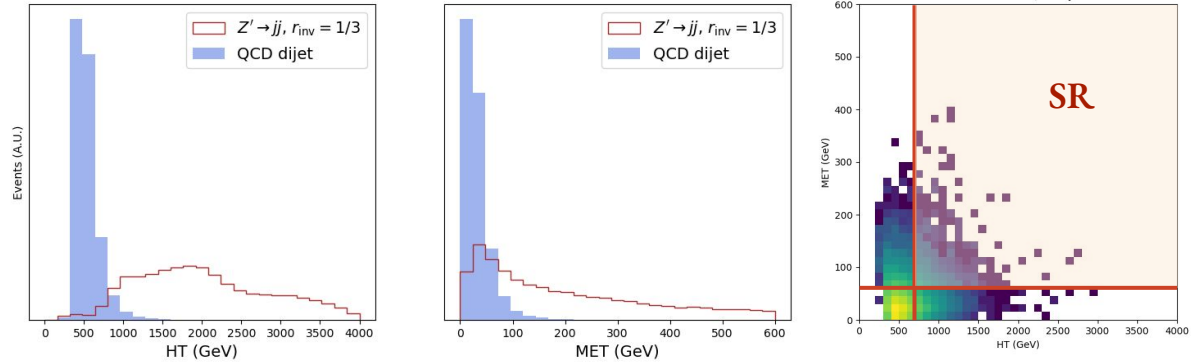


Not yet thoroughly optimized, but early indication suggest the hybrid methods are better.

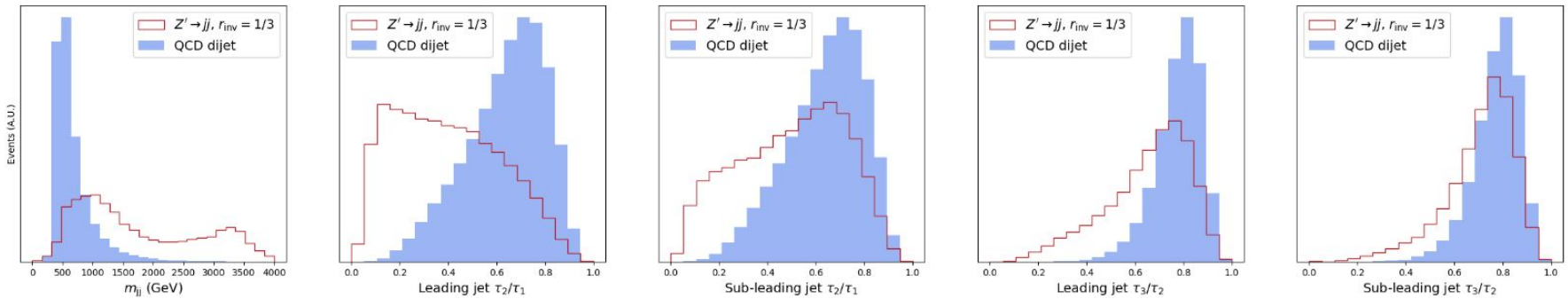
signal-to-background ratio

# Physics case

Context variables: HT & MET

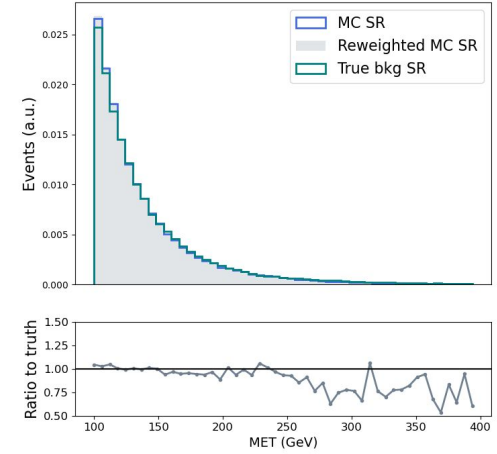
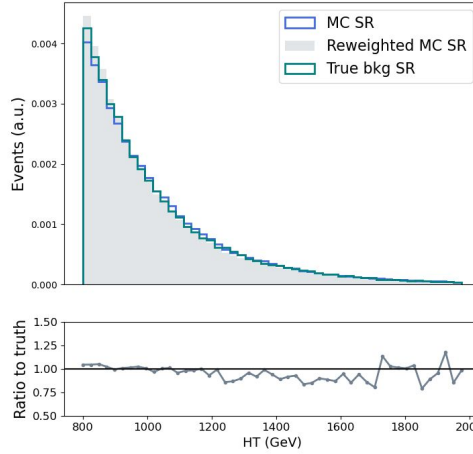


Feature variables:  $m_{jj}$  & N-subjettiness

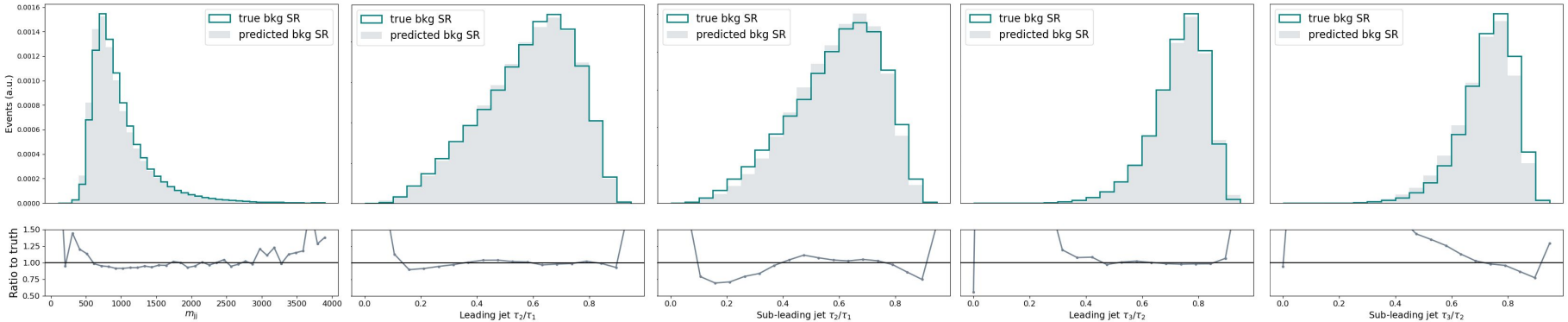


# CATHODE + SALAD hybrid

## SALAD reweighting for contexts



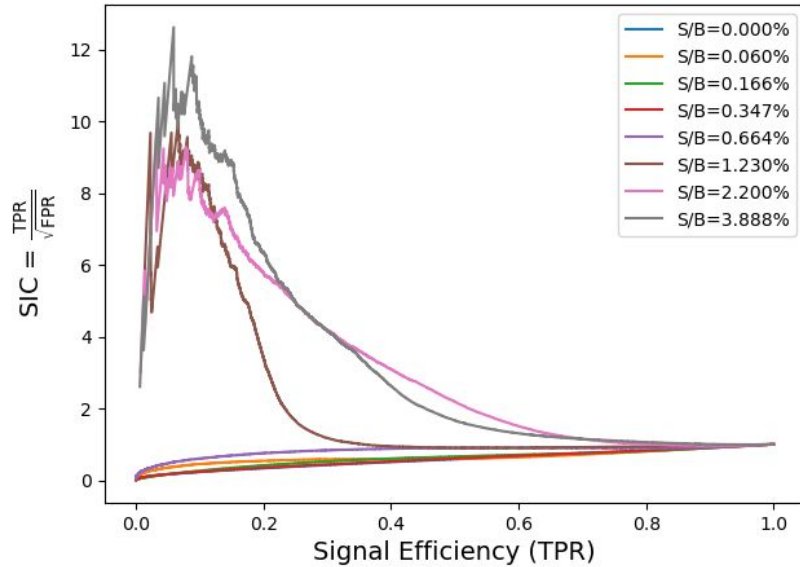
## CATHODE extrapolation for features



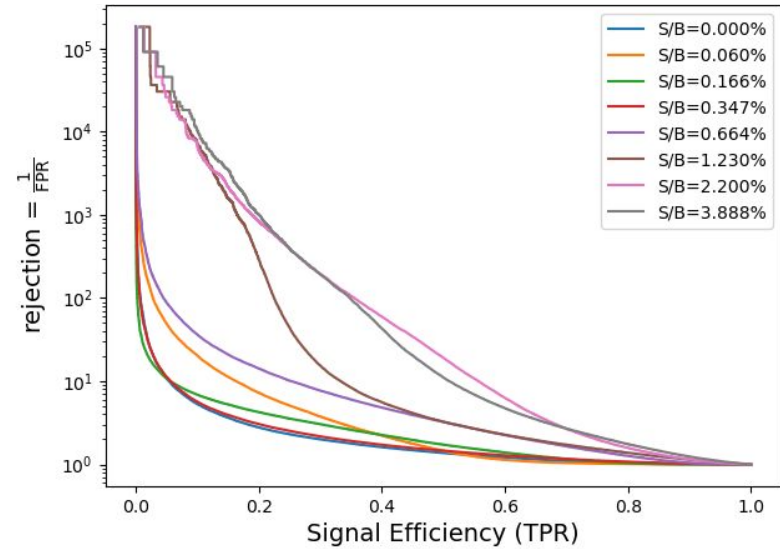
# CATHODE + SALAD hybrid

**\*\*Very preliminarily\*\*** performance plots, taken from one training

### Significance Improvement



### Rejection



# Outlook

- **This is the first case of non-resonant AD using extrapolation!**
- **There are many interesting problems associated with extrapolation that are worth exploring in the future.**
- **Stay tuned for the full results!**

**Thank you!**

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**Back up**



# SR definition in this example

SR defined to be:  $HT > 600 \text{ GeV}$  AND  $MET > 75 \text{ GeV}$

