

Combining Resonant and Tail-based Anomaly Detection

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This talk: No new models, but new results.

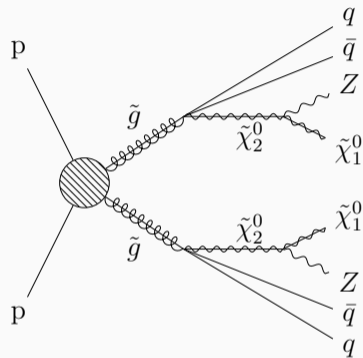
- Resonant anomaly detection and motivation
- Signal model and classical search
- Quick CATHODE recap
- Feature selection
- Results for nominal model
- Going beyond

Resonant anomaly detection and motivation

- Find feature where new physics is localised (inv. mass ...) → **Signal Region**
- Learn background template directly from data in additional features x
- Classify using difference between background template and data
→ $R(x) = \frac{p_{\text{data}}(x)}{p_{\text{bg}}(x)}$ "Anomaly score"
- Enhance significance by cutting on R
- So far: focused on localised features x e.g. τ_{ij} and global resonant feature e.g. dijet inv. mass
- Can be less strict:
 1. Can use any resonant feature as long as BG is smooth
→ Highly boosted Z,W,h .. common in decays of heavy particles
 2. Features x need not be localised
→ Anomalies on tails (e.g. $p_T^{\text{miss}}, H_T, M_{\text{eff}}$) commonly found in TeV new physics

Let's test this generalised context on a supersymmetric scenario

Signal model



- RPC SUSY scenario
- Only active sparticles : $\tilde{g}, \tilde{\chi}_1^0, \tilde{\chi}_2^0$
- $m_{\tilde{g}} - m_{\tilde{\chi}_2^0} = 50\text{GeV}$
- $m_{\tilde{\chi}_1^0} = 1\text{GeV}$
- LSP $\tilde{\chi}_1^0$ stable

→ Two highly boosted hadronic resonance at 91 GeV

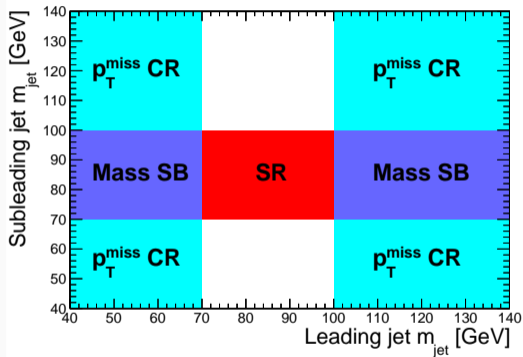
→ Large p_T^{miss} since $\tilde{\chi}_1^0$ leaves detector

→ Large H_T

Preselection

- $N_{\text{AK4 jet}} \geq 2$
- $p_T^{\text{miss}} > 300 \text{ GeV}$
- $H_T > 400 \text{ GeV}$
- at least 2 AK8 jets with $p_T > 200 \text{ GeV}$
- $|\Delta\phi_j, \vec{H}_T^{\text{miss}}| > 0.5(0.3)$ for the first 2 (up to next 2) AK4 jets, where
$$\vec{H}_T^{\text{miss}} = -\sum_{\text{AK4 jets}} \vec{p}_T$$
- no isolated photon, electron or muon candidate with $p_T > 10 \text{ GeV}$
- no isolated track with $m_T < 100 \text{ GeV}$ and $p_T > 5 \text{ GeV}$ for tracks identified as an electron/muon or else 10 GeV.

Classical search strategy



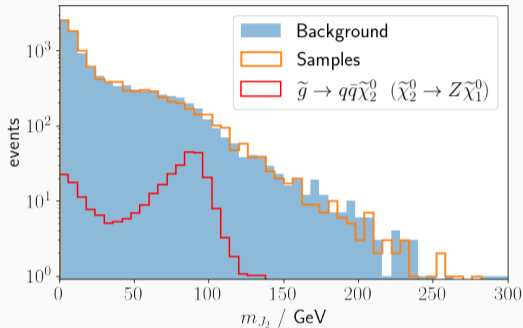
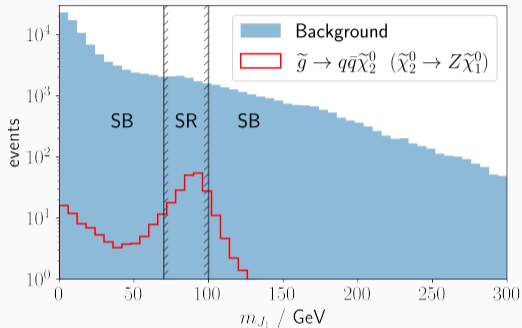
- Functional fit on leading jet m_{jet} in mass SB
→ normalisation
- p_T^{miss} -shape from p_T^{miss} -controlregion
→ shape
- Look for excess in binned p_T^{miss} -spectrum

Sirunyan et al. 2020.

Quick intro to CATHODE

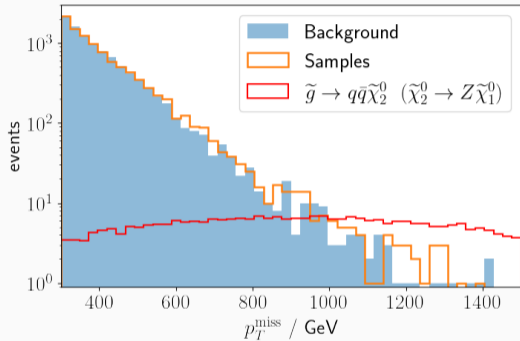
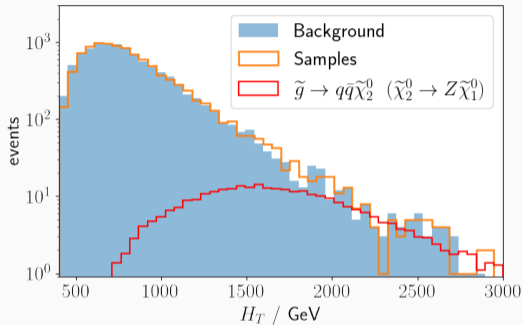
- Chose feature m where the signal is resonant and additional features x
 $\rightarrow m = m_{J1} \quad x = (m_{J2}, p_T^{\text{miss}}, H_T, \tau_{21}^{J1}, \tau_{21}^{J2})$
- Train conditional density estimator (MAF) on the sideband
- Interpolate to signal region
- Sample artificial samples (sample 40k, data 10k)
- Train classifier to classify between real data and samples: DNN with 3 layers à 64 nodes
- In practice: scan entire mass range to find the signal wherever its hiding

Distribution of features



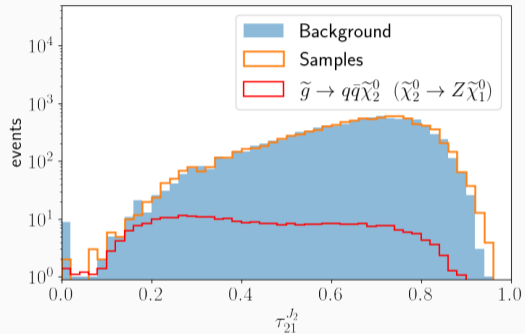
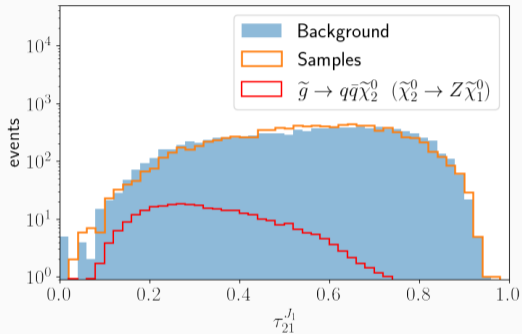
- Chose SR in $m_{J_1} \in [m(1 - \frac{4}{3}\sigma_m), m(1 + \frac{2}{3}\sigma_m)]$ with $\sigma_m = 15\%$
→ SR of classical search for boosted Z

Distribution of features



- Both H_T and p_T^{miss} are useful classification features
- Live on the tails

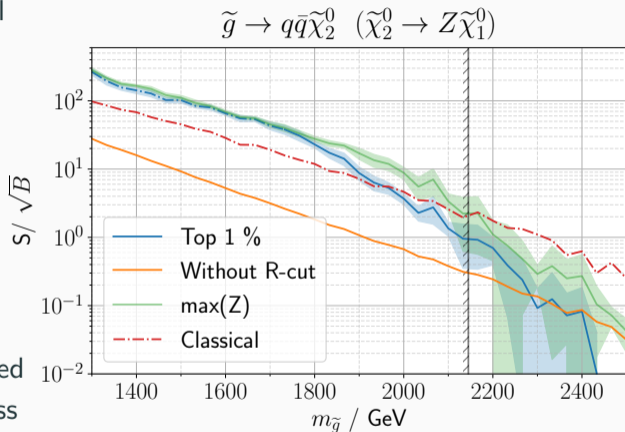
Distribution of features



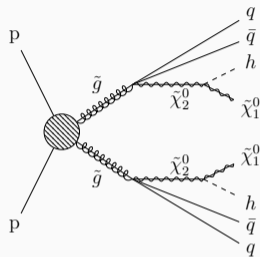
- Still useful: S/\sqrt{B} from ~ 34 to ~ 40 for $m_{\tilde{g}} = 1700$ GeV

- Leave one out cross validation for DE and classifier
- Ensemble 4 x 10 DEs for sampling
- Assign avg. anomaly score of 10 classifiers
- Reshuffle 5x and repeat
- Repeat 10x with independent signal events
→ Expected sensitivity
- Focus on S/\sqrt{B} as a metric in place of a proper statistical analysis

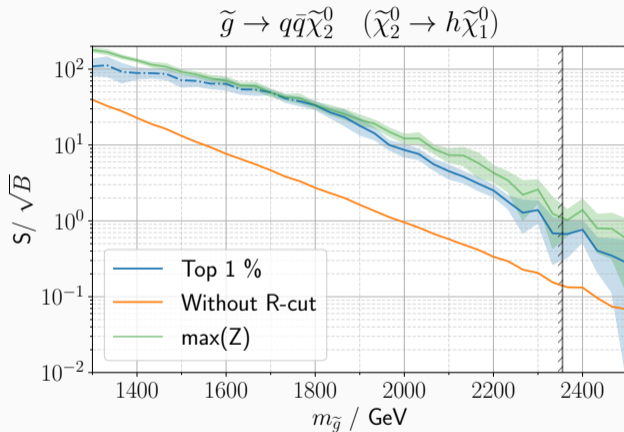
- Cant directly compare classical (multibin) search with S/\sqrt{B} \rightarrow compare to S/\sqrt{B} for $p_T^{\text{miss}} > 800\text{GeV}$ red
- Top 1% events sorted by anomaly score blue
- Maximise S/\sqrt{B} with $B \geq 5$ green
- Vertical line: 300 fb^{-1} recreated expected 95%CL excluded mass
- 1% metric only slightly worse!



125 GeV Higgs

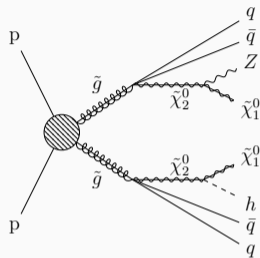


- No b-tag necessary

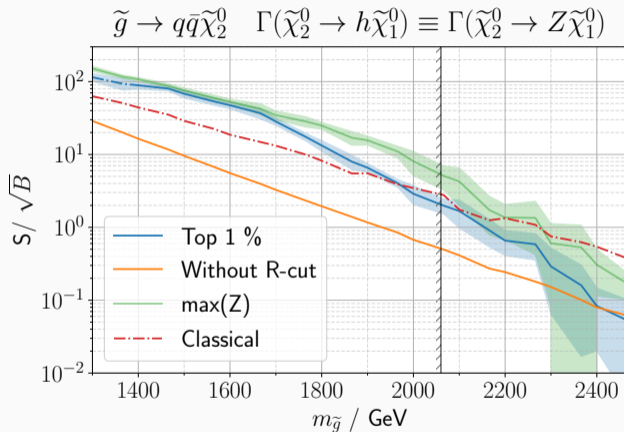


Tumasyan et al. 2022, @ 137/fb.

125 GeV Higgs and Z



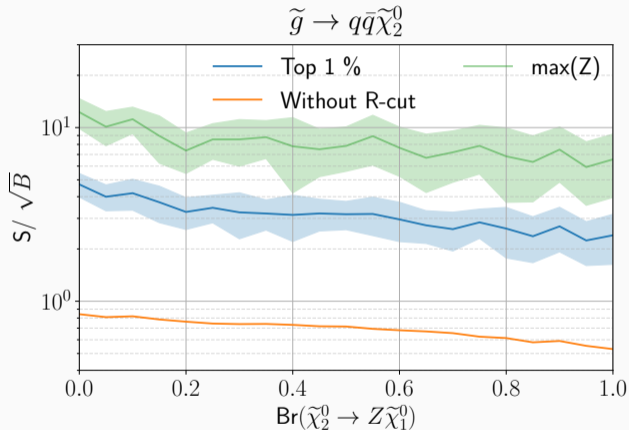
- $m_{J1} \in [70 \text{ GeV}, 140 \text{ GeV}]$



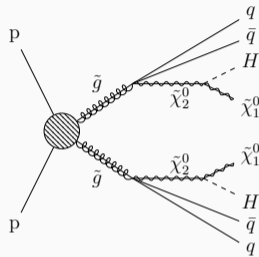
Sirunyan et al. 2018, @ 35.9/fb.

125 GeV Higgs and Z

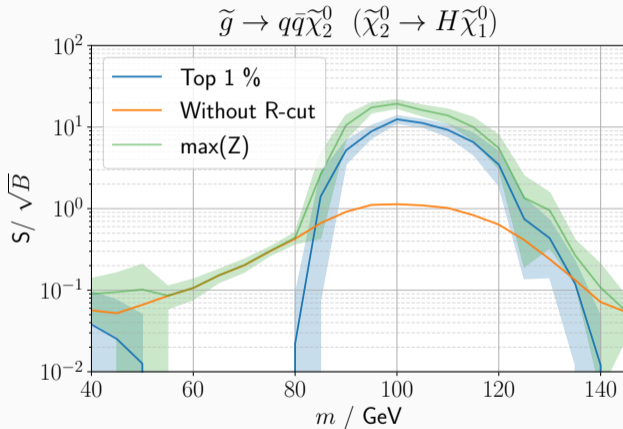
- $m_{\tilde{g}} = 2000$ GeV
- Almost doesn't care about branching ratio



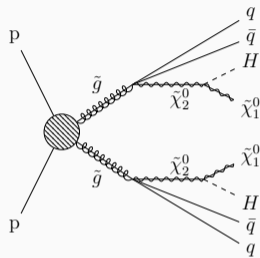
Additional Higgs



- $m_{\tilde{g}} = 2000 \text{ GeV}$
- $m_H = 100 \text{ GeV}$

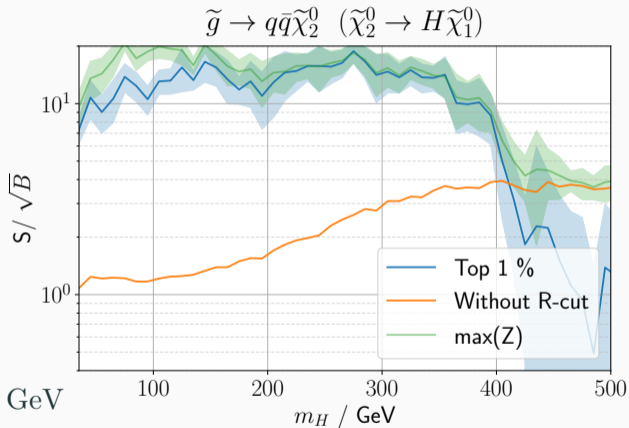


Additional Higgs







- $m_{\tilde{g}} = 2000$ GeV

- Enhancement up to $m_H \sim 400$ GeV



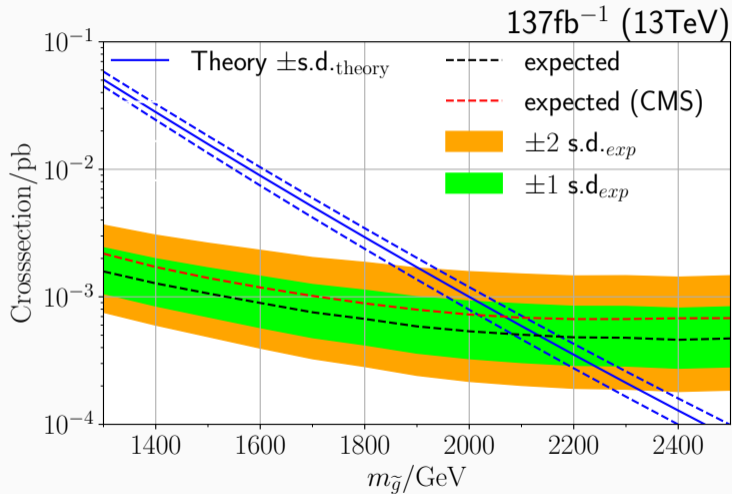
- CATHODE works even for features on tails of distributions
- We don't lose much by going model agnostic compared to dedicated search
- Covers multiple signal models at once without extra work

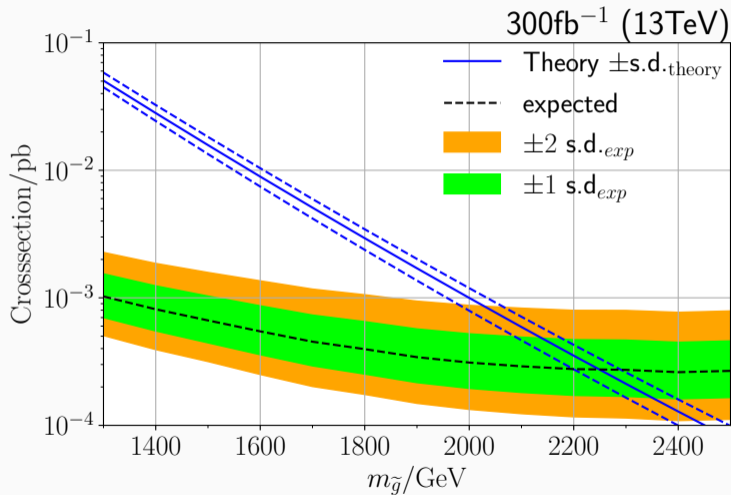
References

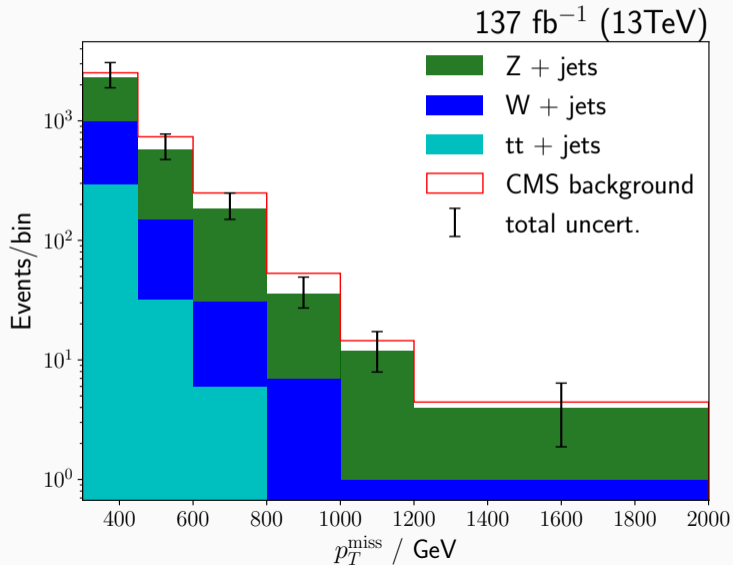
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Backup







Oversample

