

HEPData for the ATLAS Dark Jet Resonance Search

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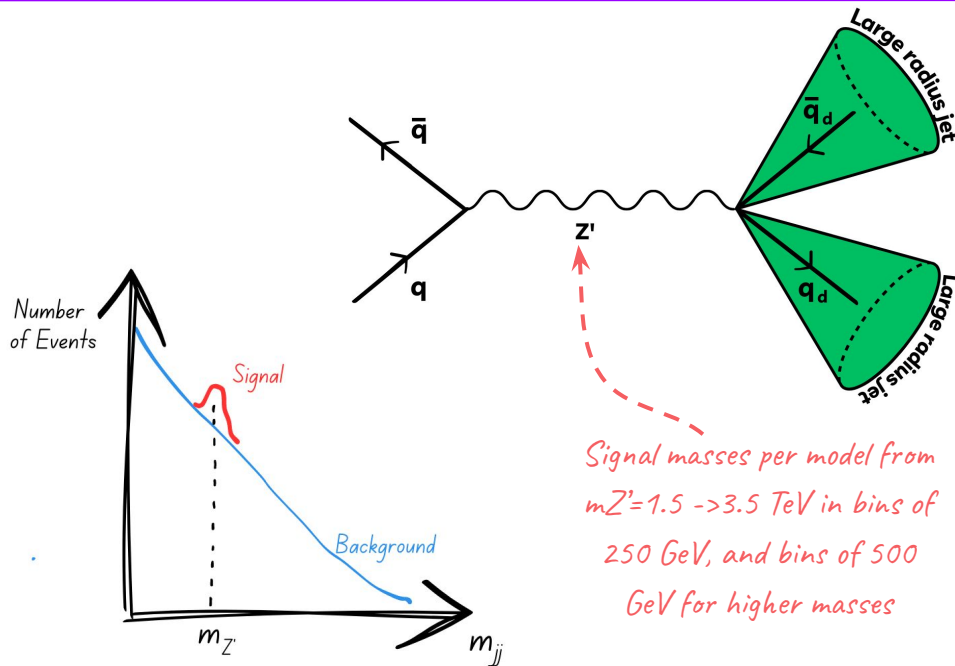
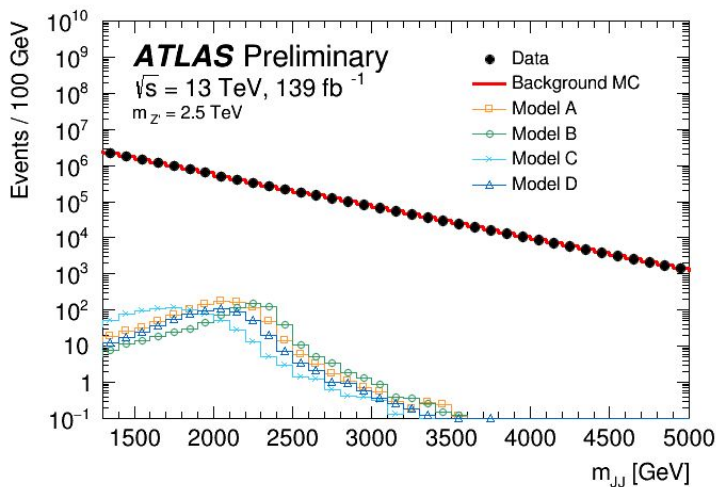
Dark Jet Resonance Search Analysis Strategy:

See [Dilia's talk](#) for more information!

- Reconstructed Objects: Two Large radius jets (Trimmed LCTopo jets)
- Trigger on Large-R jets
- Preselection cuts: $|\eta(j_{1,2})| < 2.0$ $p_T(j_1) > 500$ GeV, $p_T(j_2) > 400$ GeV $m(j_{1,2}) > 50$ GeV $m_{JJ} > 1.3$ TeV

Signal scenario: Two dark jets

- 1) Tag them using substructure information
- 2) Look for a resonance over dijet background



ATLAS HEPData Recommendations: Statistically Limited Searches

- **Preservation of analysis logic:**
 - cutflow table for each signal region, for several reproducible or well-defined signal samples
- **Preservation of numerical values:**
 - central values for the search regions in each bin
- **Systematic uncertainties/correlations:**
 - A simple stat/syst breakdown is sufficient if the analysis is dominated by statistical uncertainties
- **SM background predictions**
- **Limits**
 - Exclusion limits + necessary information for a theorist to emulate the analysis

Di-jet invariant mass distribution

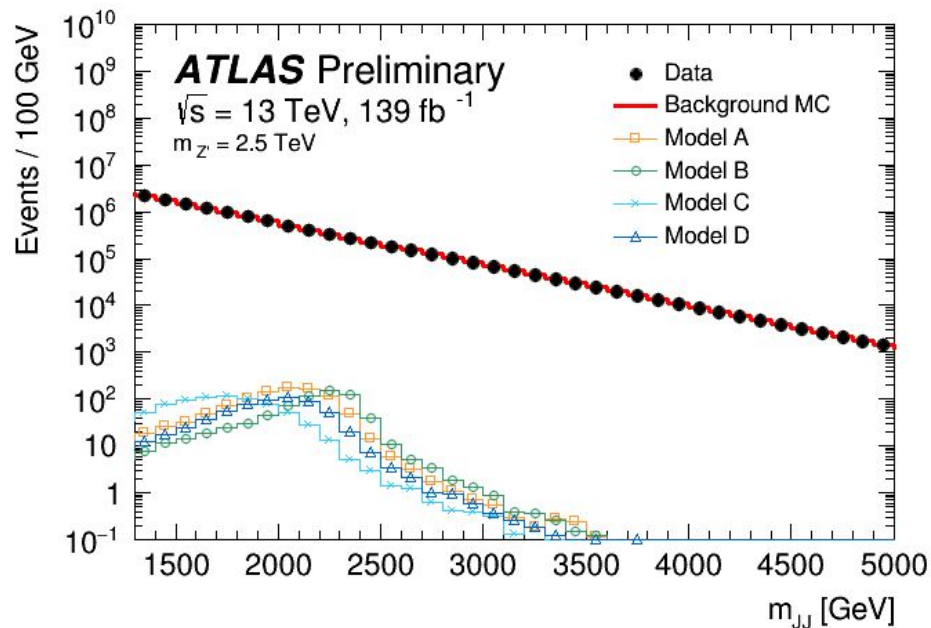
- Data, signal, background counts for m_{jj} in preselection
- Signal m_{jj} shape before signal region cuts for reinterpretation validation
- Normalisation:
 - simulated background is normalised to the data
 - signals are normalised to a production cross-section of 10 fb.

preselection

$$m_{JJ} > 1.3 \text{ TeV}$$

$$m(j_{1,2}) > 50 \text{ GeV} \quad |\eta(j_{1,2})| < 2.0$$

$$p_T(j_1) > 500 \text{ GeV}, p_T(j_2) > 400 \text{ GeV}$$



Distributions of the number of tracks associated with the two leading jets

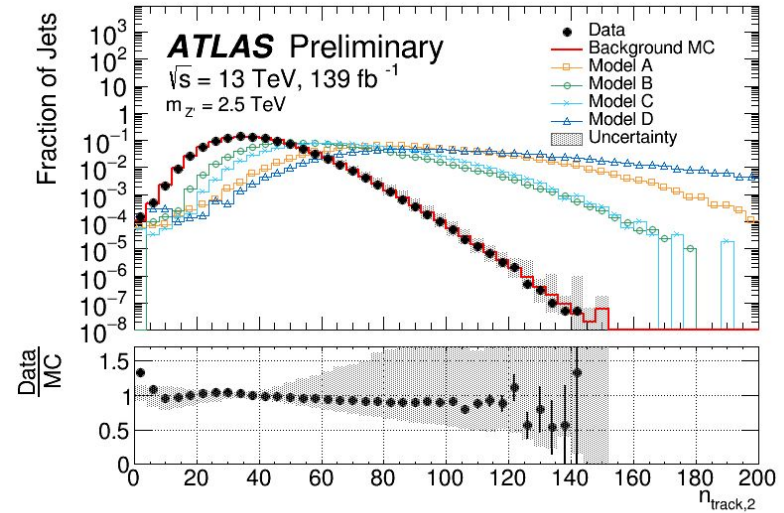
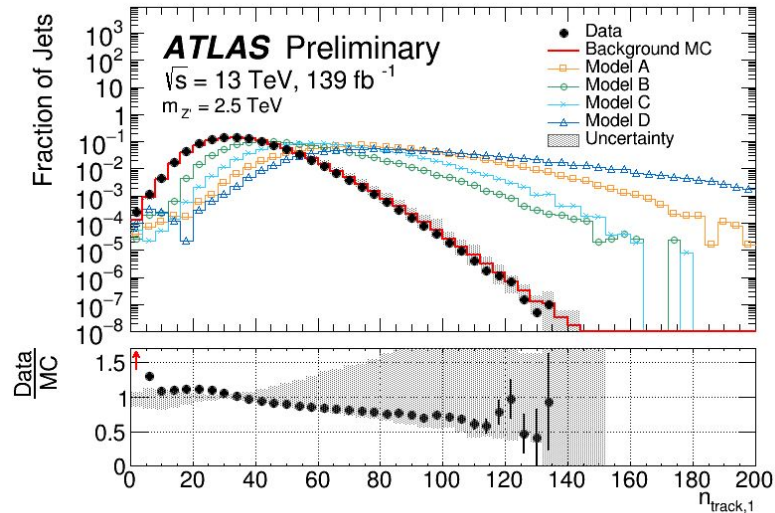
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- The number of tracks matched to the untrimmed jets via ghost association
- Uncertainty band around background: modelling uncertainty

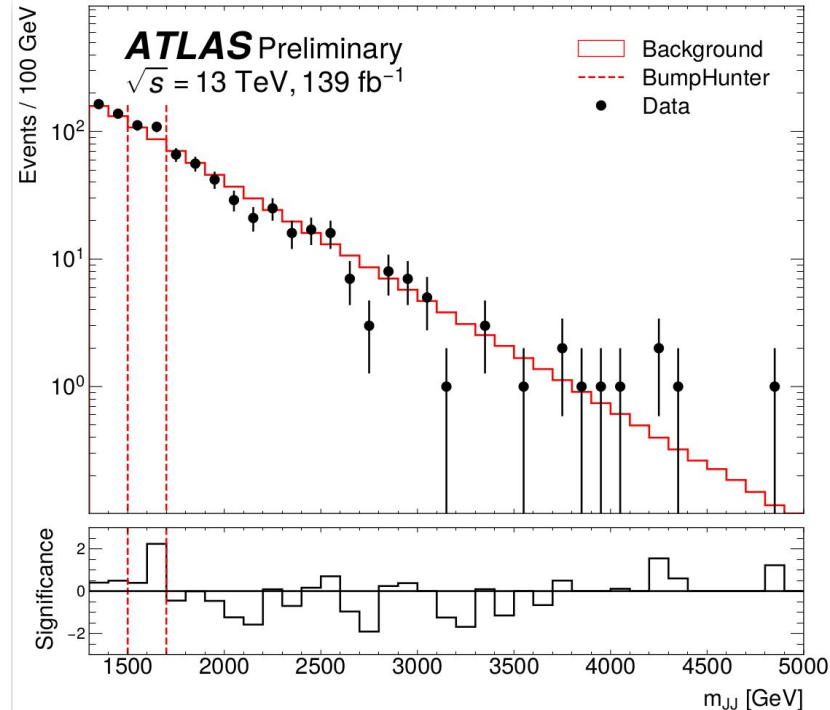


Model Independent Limits

- Bump hunter test performed to find localised excess in model-independent manner
- Algorithm scans the distributions using a sliding window of variable width between two bins and half the total number of bins.
- Highest excess: between $1500 < m_{jj} < 1700$ GeV, with a global p-value of 0.63
- Data / background yield for m_{jj} in the SR
 - + statistical error on data

Signal Region (SR)

$$n_{\text{track},1}^{\epsilon} > 0 \text{ and } n_{\text{track},2}^{\epsilon} > 0$$

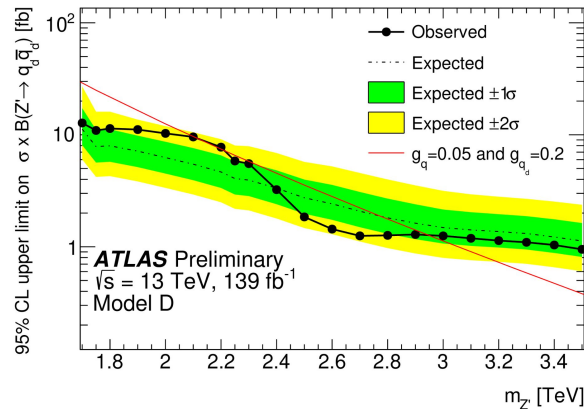
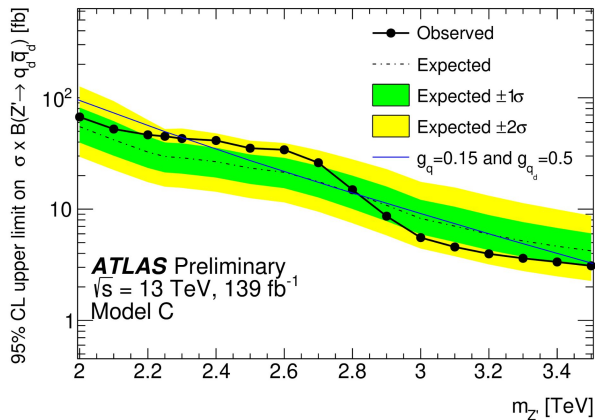
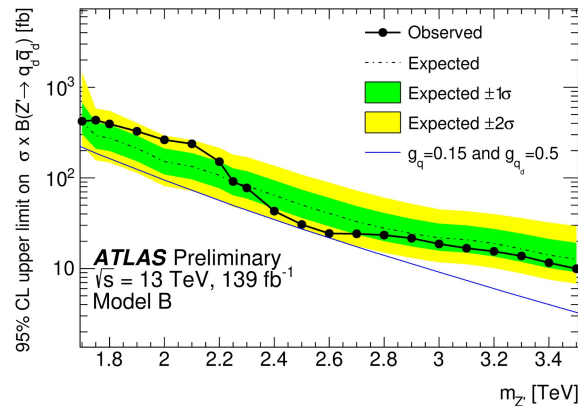
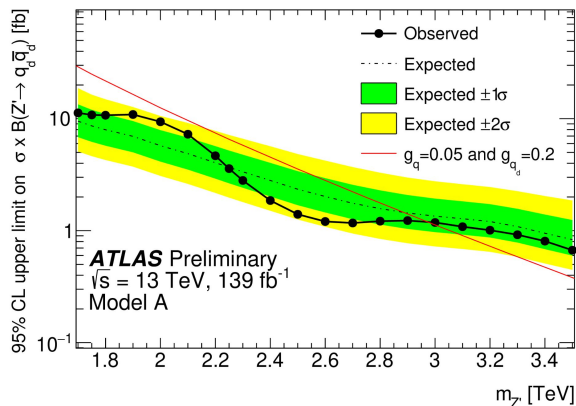


Model Dependent Limits

Signal Region (SR)

$$n_{\text{track},1}^{\epsilon} > 0 \text{ and } n_{\text{track},2}^{\epsilon} > 0$$

- Expect limits
- Observed limits
- $\pm 1\sigma$ and $\pm 2\sigma$ uncertainty ranges
- Look into storing likelihoods as **pyhf** objects
- Theory predictions typically not provided for HEPData



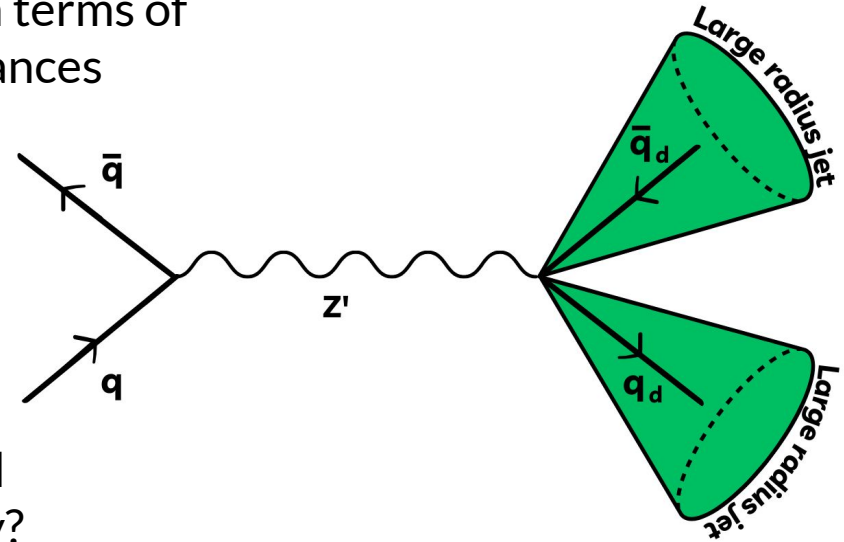
Cutflow Table : Relative Signal Efficiencies

Selection / Model	A	B	C	D
$m_{JJ} > 1.3 \text{ TeV}$	92.9	94.8	80.9	91.8
Jet trigger	93.0	93.2	92.5	92.3
$m_{J_{1,2}} > 50 \text{ GeV}, p_{T,J_1} > 500 \text{ GeV}, p_{T,J_2} > 400 \text{ GeV}$	88.5	60.0	81.3	56.1
$ \eta_{J_{1,2}} < 2$	99.9	99.9	100	100
$m_{J_{1,2}} < 600 \text{ GeV}, p_{T,J_{1,2}} < 3000 \text{ GeV}$	99.8	99.7	99.9	99.8
Signal Region ($n_{\text{track},1}^e > 0$ and $n_{\text{track},2}^e > 0$)	37.0	2.7	11.6	55.5

Table 3: Relative efficiency (in %) of each analysis selection with respect to the previous one for models A through D and $m'_Z = 2.5 \text{ TeV}$. Approximately 90000 events were generated per signal.

Discussion Points

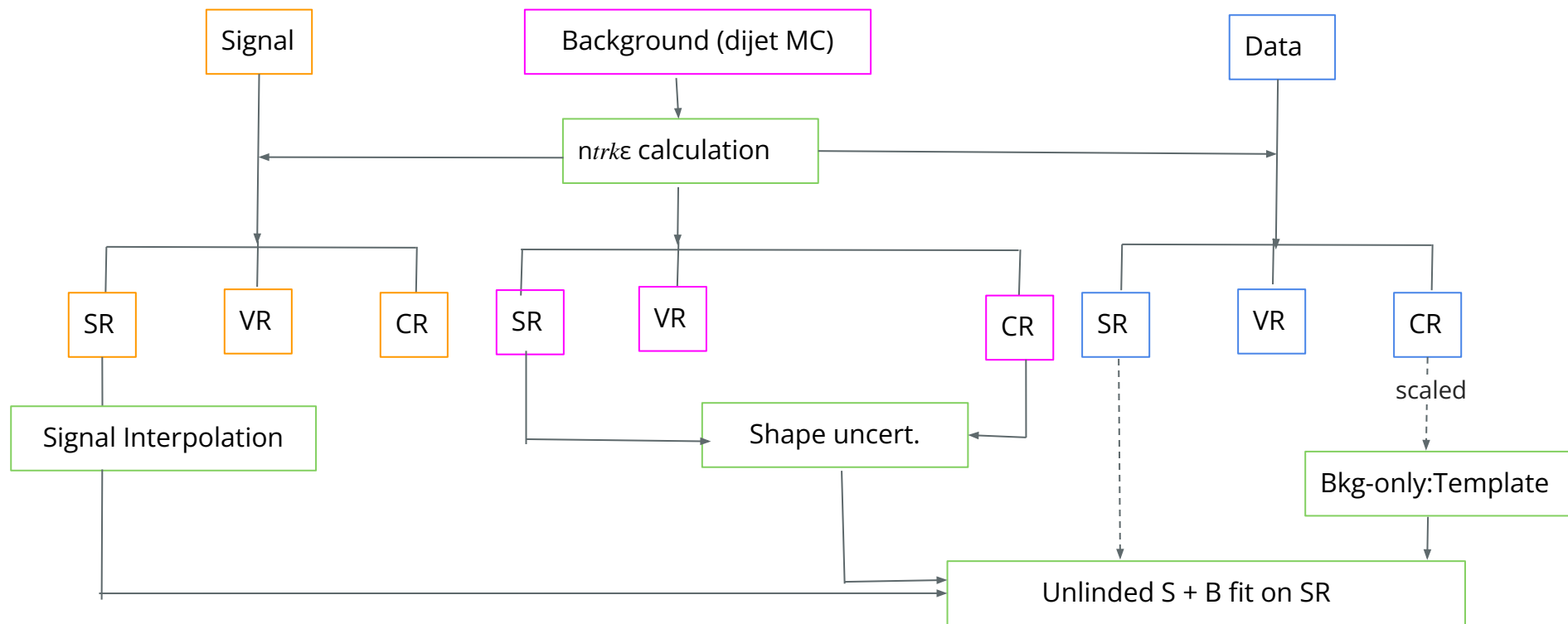
- 4 benchmark models considered: by no means an exhaustive selection of dark QCD scenarios!
 - First time interpreting ATLAS data in terms of promptly-decaying dark QCD resonances



- **Theorists:** What other information would be useful to include in our HEPData entry?
 - We would love to maximize the re-interpretability of our results.

Backup

Analysis workflow

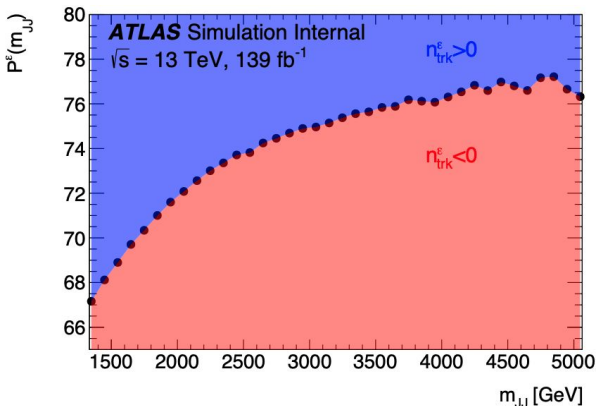
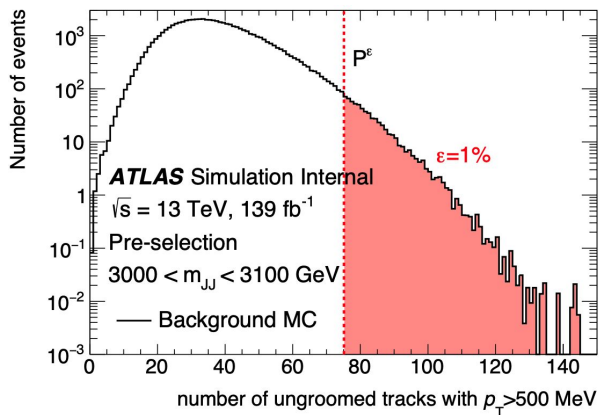


SR: $ntrkE(j1) > 0$ and $ntrkE(j2) > 0$
CR: $ntrkE(j1) < 0$ and $ntrkE(j2) < 0$
VR: $ntrkE(j1) > 0$ and $ntrkE(j2) < 0$

VR & CR scaled to expected SR when testing the fitting

Decorrelation with n_{trk}^ϵ

1. In bins of m_{JJ} , evaluate the cut value on n_{trk} that corresponds to a given target background efficiency ($\epsilon=1\%$)

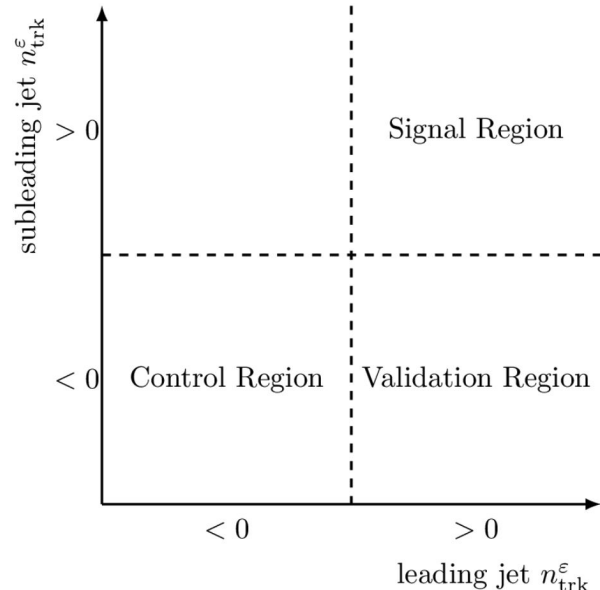


2. Fit this profile in order to get a smooth distribution of expected n_{trk} percentiles $\rightarrow P^\epsilon$

3. Define n_{trk}^ϵ :

$$n_{\text{trk}}^\epsilon(m_{\text{JJ}}) = n_{\text{trk}} - P^\epsilon(m_{\text{JJ}})$$

analysis regions



Percentiles are defined from full MC *, since ..

- Defining from full data => unblinding the SR
- Defining from data CR would change the shape of the m_{JJ} distribution

* n_{trk} data monte carlo differences are covered with systematics (more info [backup slides](#))

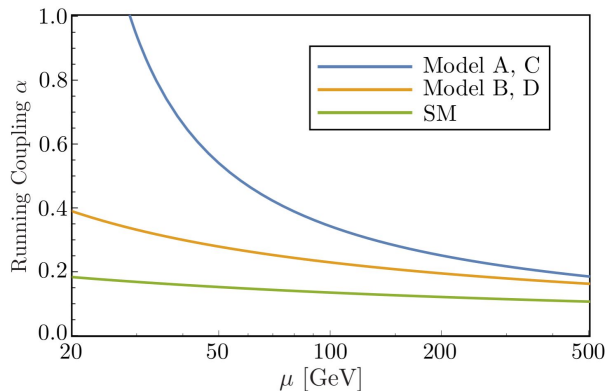
Target Model

Unusual dijet signatures that could arise from a QCD-like dark sector.

- Assume a heavy mediator Z' links dark sector and visible sector

[arXiv:1712.09279]

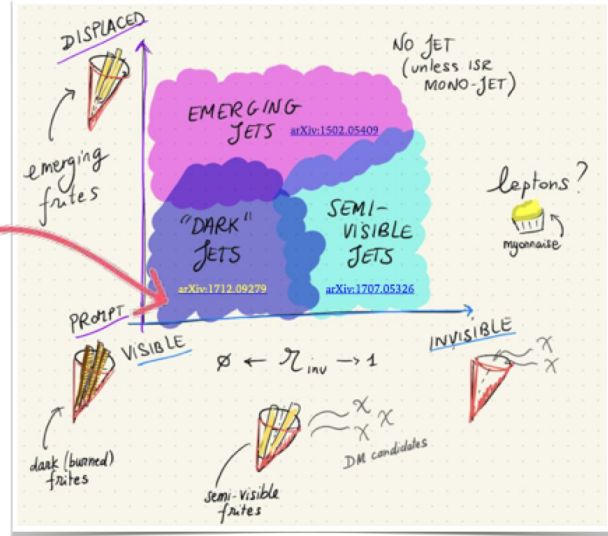
	N_d	n_f	Λ_d (GeV)	$\tilde{m}_{q'}$ (GeV)	m_{π_d} (GeV)	m_{ρ_d} (GeV)
A	3	2	15	20	10	50
B	3	6	2	2	2	4.67
C	3	2	15	20	10	50
D	3	6	2	2	2	4.67



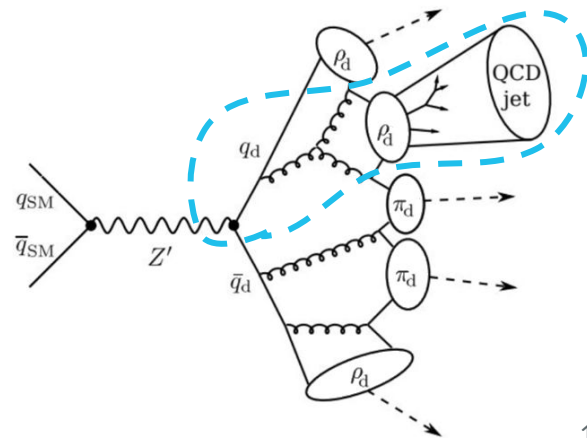
Consider **4 benchmark models..**

Compared to the SM, the dark jets have:

- stronger running couplings
- more soft particles (\Rightarrow more tracks)
- larger jets, due to **double hadronization**

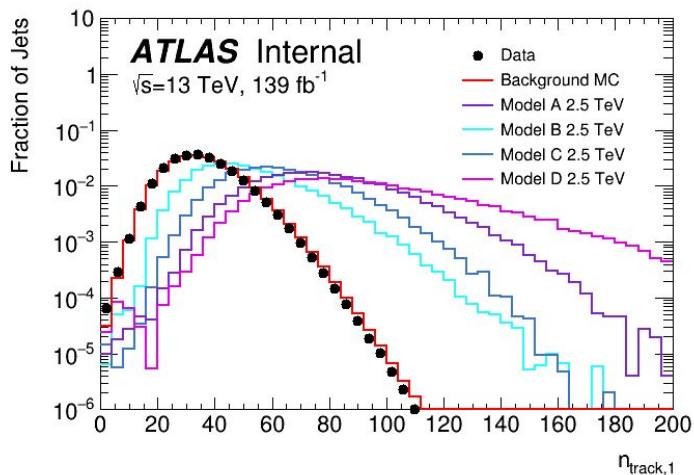


Graphics by: C. Doglioni, K. Pedro

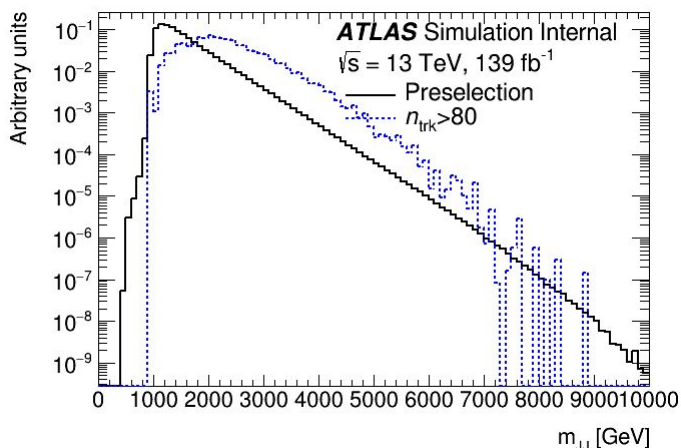


Tagging Dark Jets

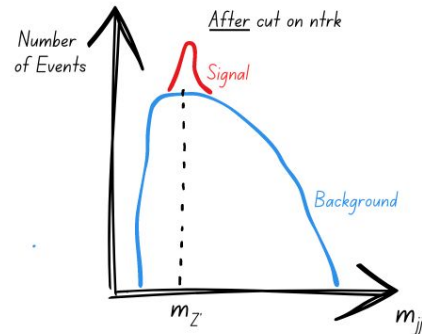
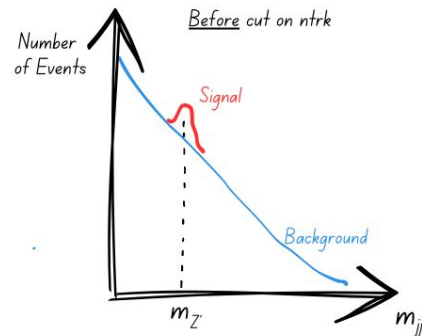
The ungroomed number of tracks inside the jet, n_{trk} , is a strong discriminating variable.



But a sharp n_{trk} cut significantly sculpts the background dijet mass spectrum



Which impacts our bump hunt!



Solution:

Define a new key discriminator n_{trk}^ϵ decorrelated from the dijet mass

Preselection

Triggers:

HLT_j460_a10_sub_L1J100 for 2015 and 2016

HLT_j460_a10t_lcw_jes_L1J100 for 2017 and 2018

Preselection cuts: [\(Twiki\)](#)

Two $R = 1.0$ **trimmed** LCTopo jets

$$|\eta(j_{1,2})| < 2.0$$

$$p_T(j_1) > 500 \text{ GeV}, p_T(j_2) > 400 \text{ GeV}$$

$$m(j_{1,2}) > 50 \text{ GeV}$$

$$p_T(j_{1,2}) < 3000 \text{ GeV}$$

$$m(j_{1,2}) < 600 \text{ GeV}$$

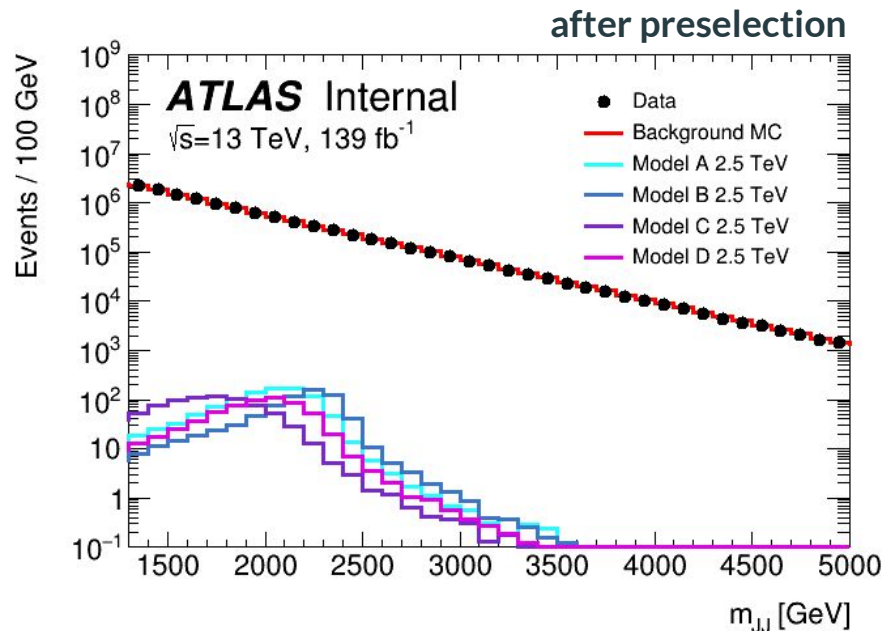
For background Monte Carlo (di-jet quality cut):

$$0.7 < \frac{p_T^{\text{leading jet}} + p_T^{\text{subleading jet}}}{2 \times p_T^{\text{leading truth jet}}} < 1.3$$

Derivation: EXOT3

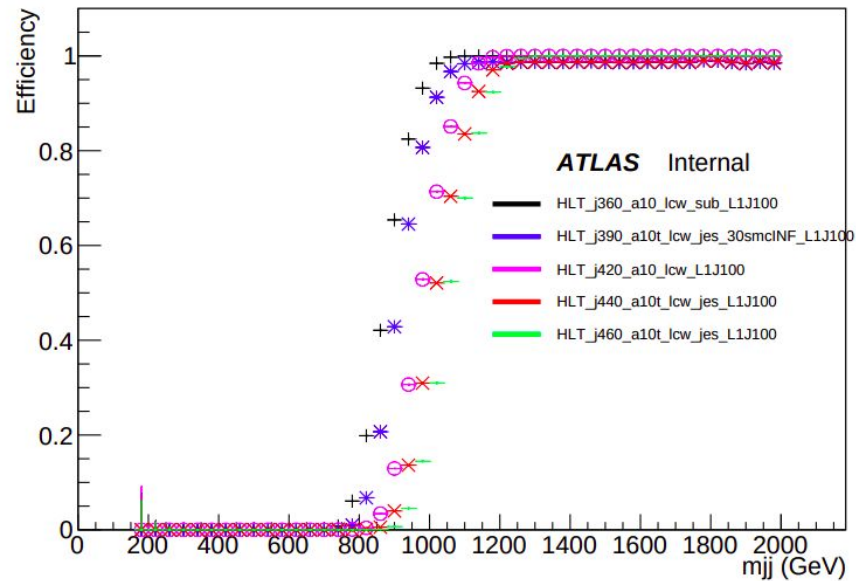
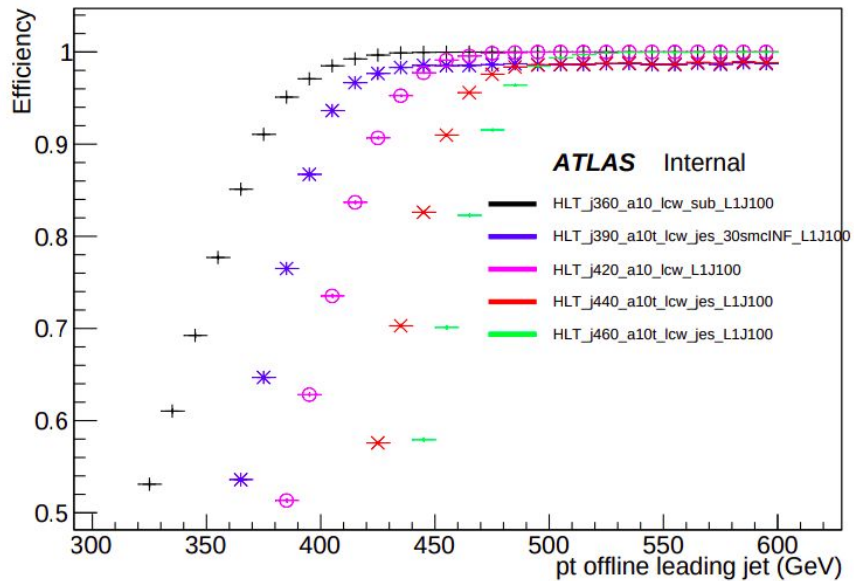
Analysis framework: [xAODAnaHelpers](#).

Useful links: [Glance](#), [Analysis code](#), [Internal note](#).



- signal distributions are:
 - relatively broad
 - shifted to lower values than Z' mass
- no **trimming** => narrower signal peak BUT would increase background significantly

Trigger Turn Ons

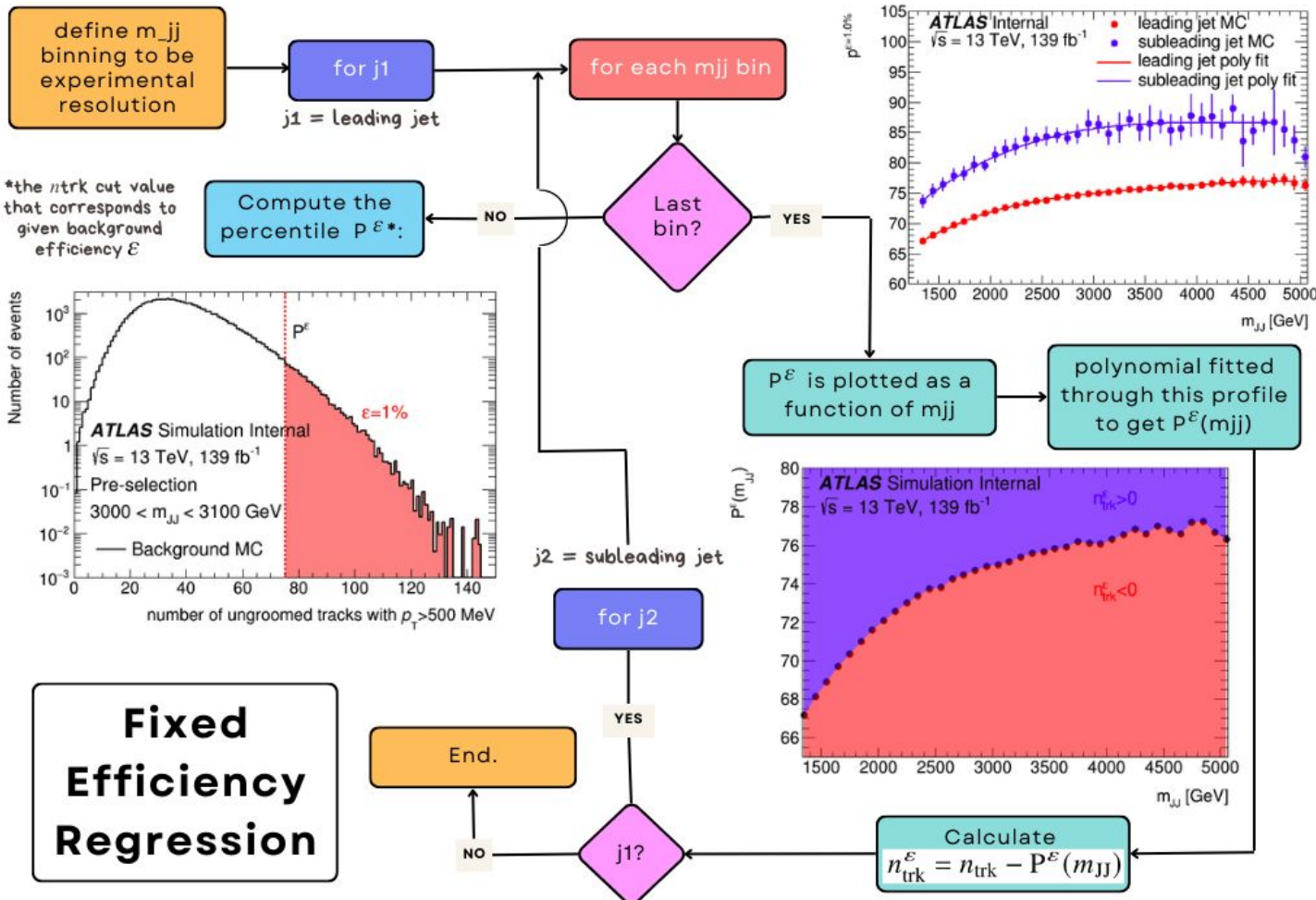


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n_{trk} data monte carlo differences are covered with systematics



Make it linear with 3 plots...