

Model-agnostic search for dijet resonances with the CMS detector

Results to appear on
CDS [\[link\]](#)

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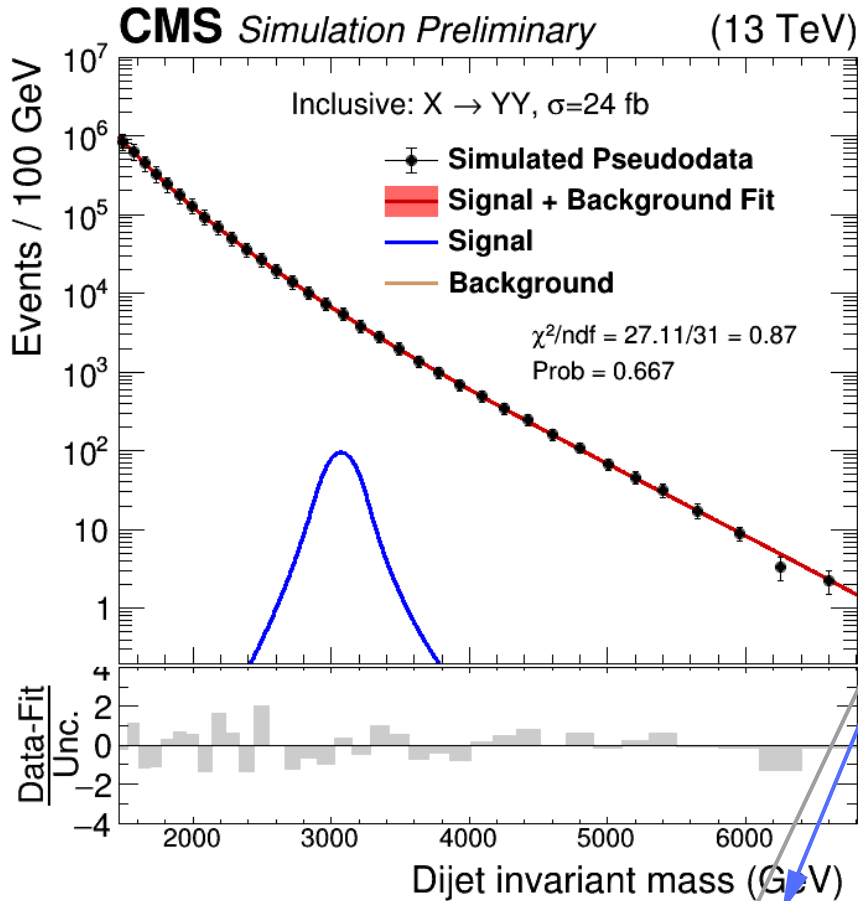
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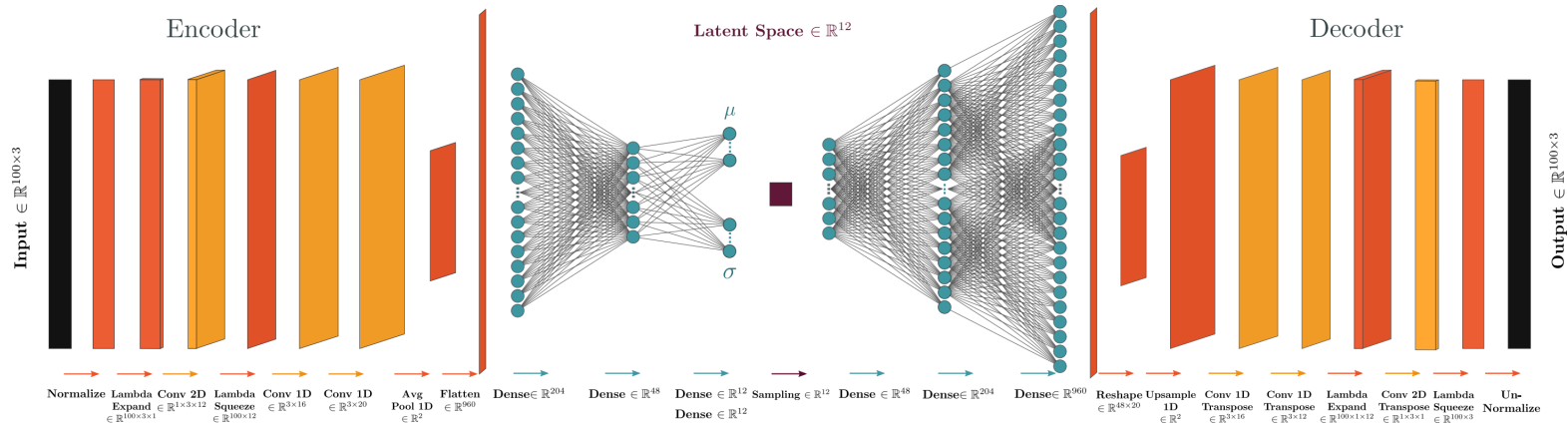
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Overview

- Dijet events (anti- k_T , $R = 0.8$)
- Look for a narrow resonance
- Leverage substructure
- Use state-of-the-art AD methods (as of two years ago)
- Compare **5** techniques for anomaly detection



Methods: Variational Autoencoder



- Encodes up to 100 PF* constituents per jet
- Trained with jets from a QCD-dominated sideband ($\Delta\eta > 1.4$)
- Final score: lowest reconstruction loss of the two jets
- Background sculpting controlled with quantile regression

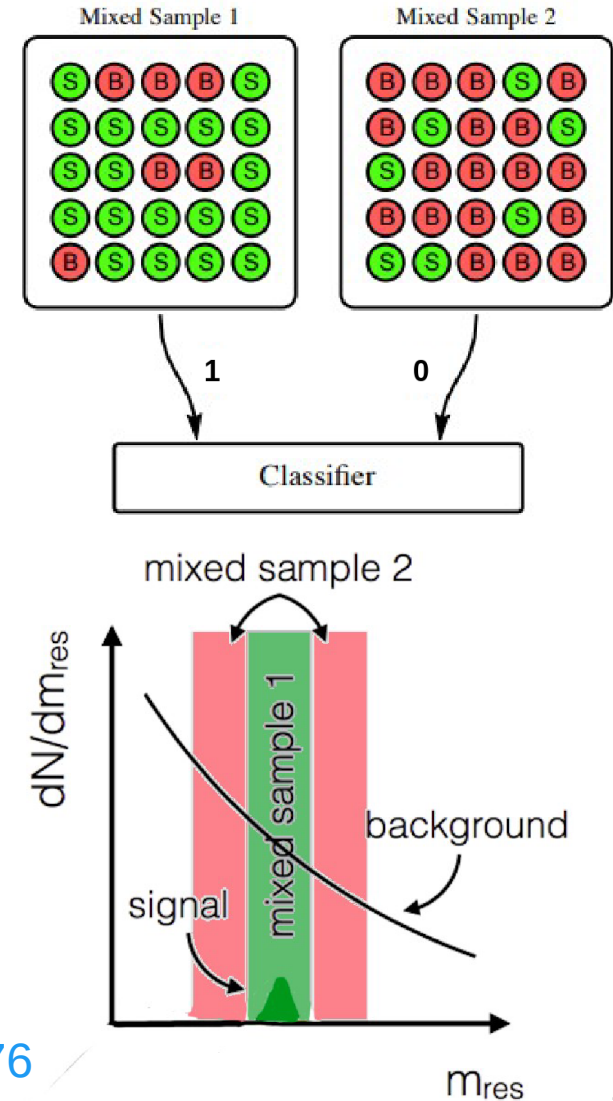
Methods: Weak Supervision

Train a classifier between data and a background-like sample

- **CWoLa**: background taken from sidebands
- **CATHODE**: background interpolated from sidebands
- **Tag N' Train**: autoencoder preselection, targets events with two anomalous jets

Fewer features for CATHODE than CWoLa/TNT

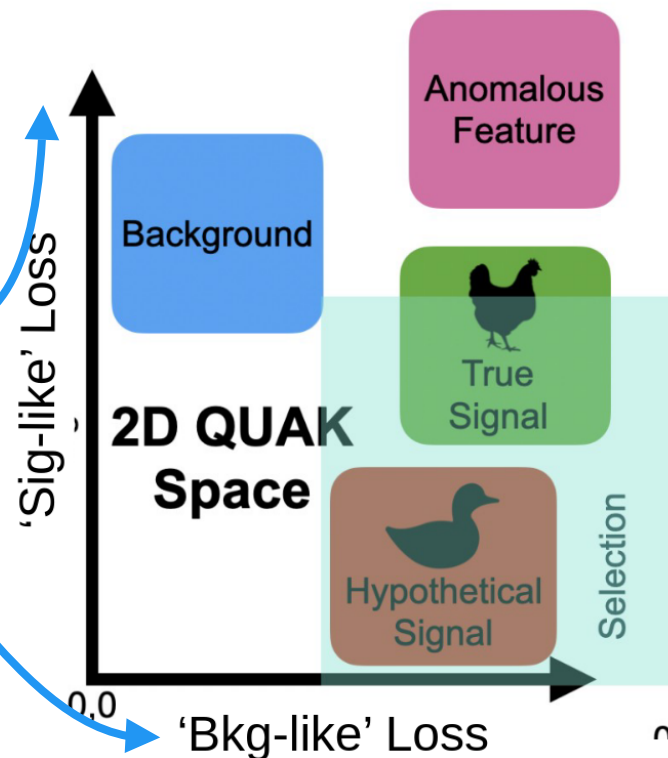
CWoLa: [1902.02634](#) / CATHODE: [2109.00546](#) / TNT: [2002.12376](#)



Methods: QUAK

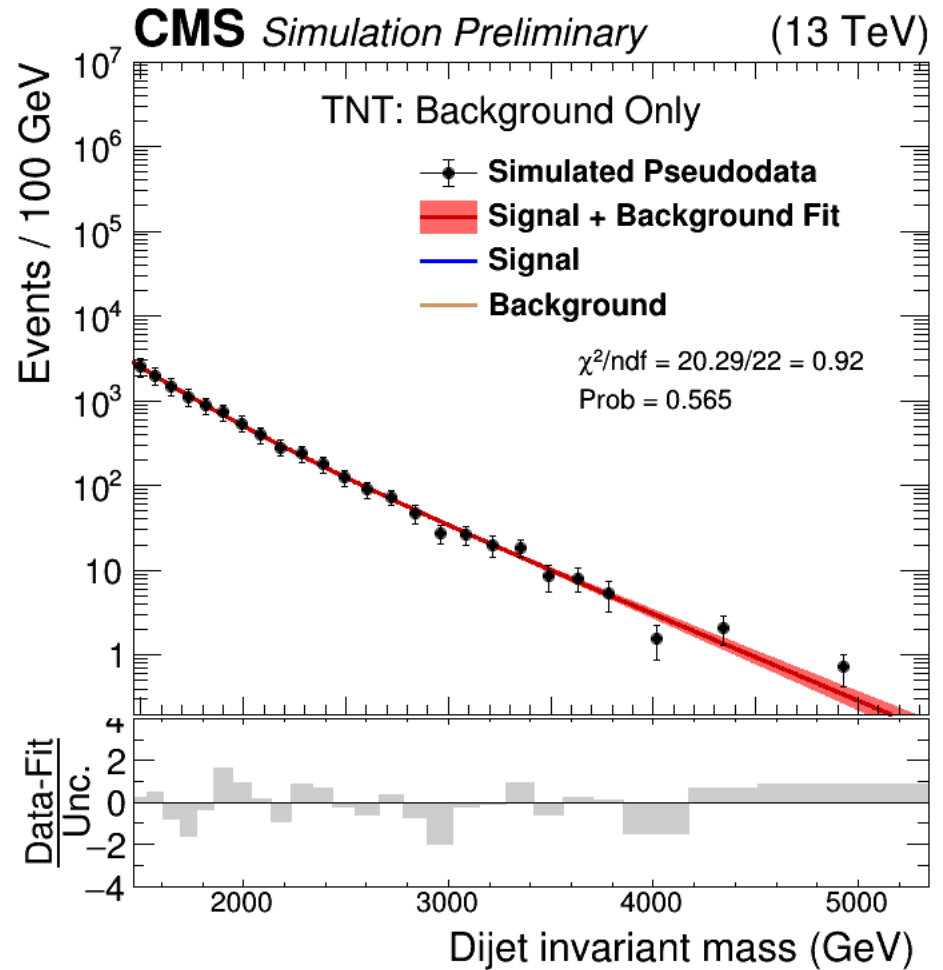
- Hybrid approach, encoding a **prior** on signal-like features
- Train two normalizing flows:
 - On a mixture of signal MCs
 - On background MC
- The losses define a **2D QUAK space**
- The signal is somewhere in that space...

Hypothetical QUAK Space



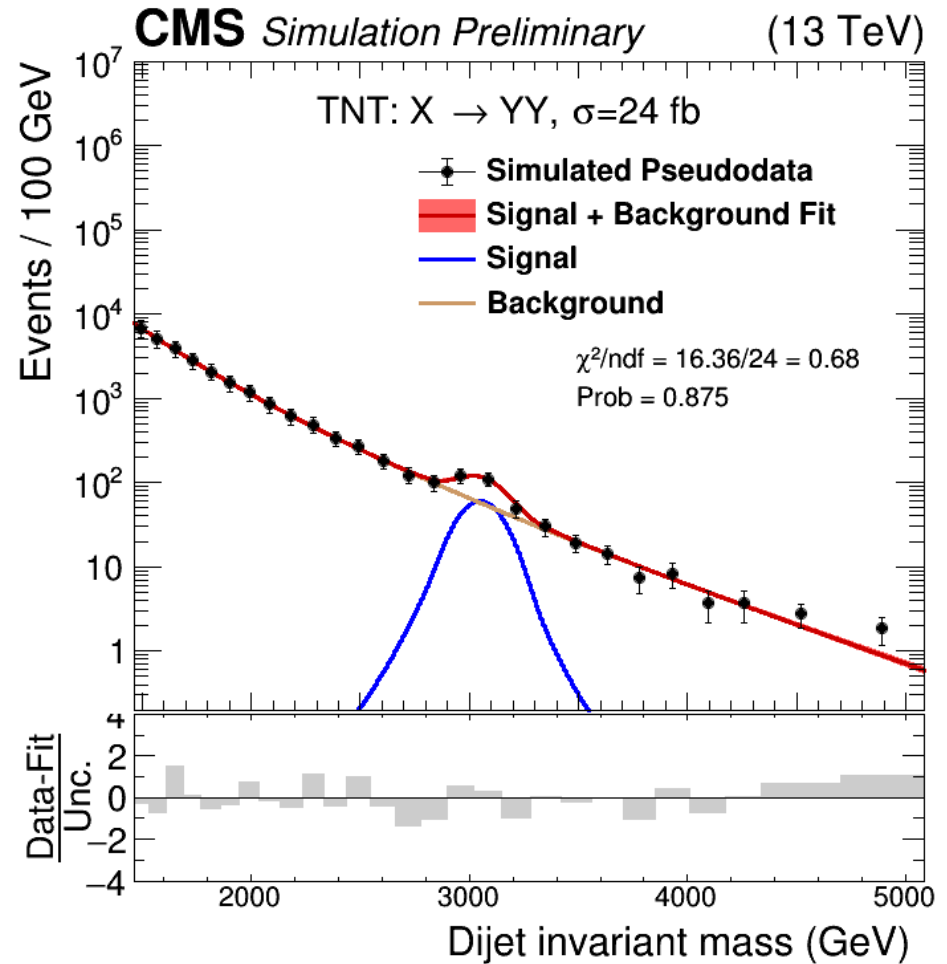
After Tagging

- Choose a working point
- Select events
- Look at m_{jj} spectrum
- Fit with analytic functions



After Tagging

- Choose a working point
- Select events
- Look at m_{jj} spectrum
- Fit with analytic functions
- Find a bump (maybe)
- Derive a p -value



We use everything

Compression	VAE/TNT
Regression	QR for VAE
Classification	CWoLa et al.
Density estimation	CATHODE/QUAK
Event generation	CATHODE
Hand-crafted analytical function	Final m_{jj} fit

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... except data ...

Complementarity

Do all methods find
the same events?

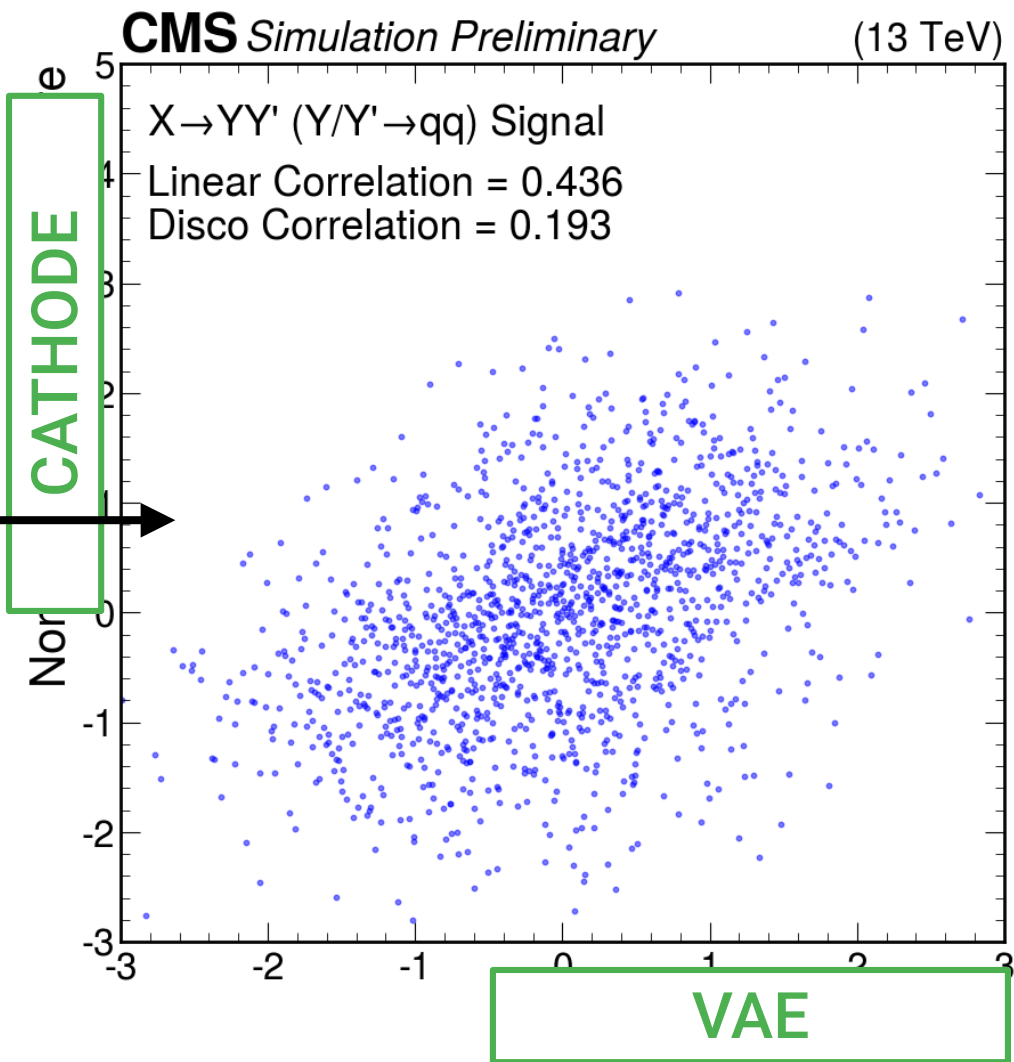
Complementarity

Do all methods find the same events?

Check correlation between scores

≈ LHCO R&D dataset

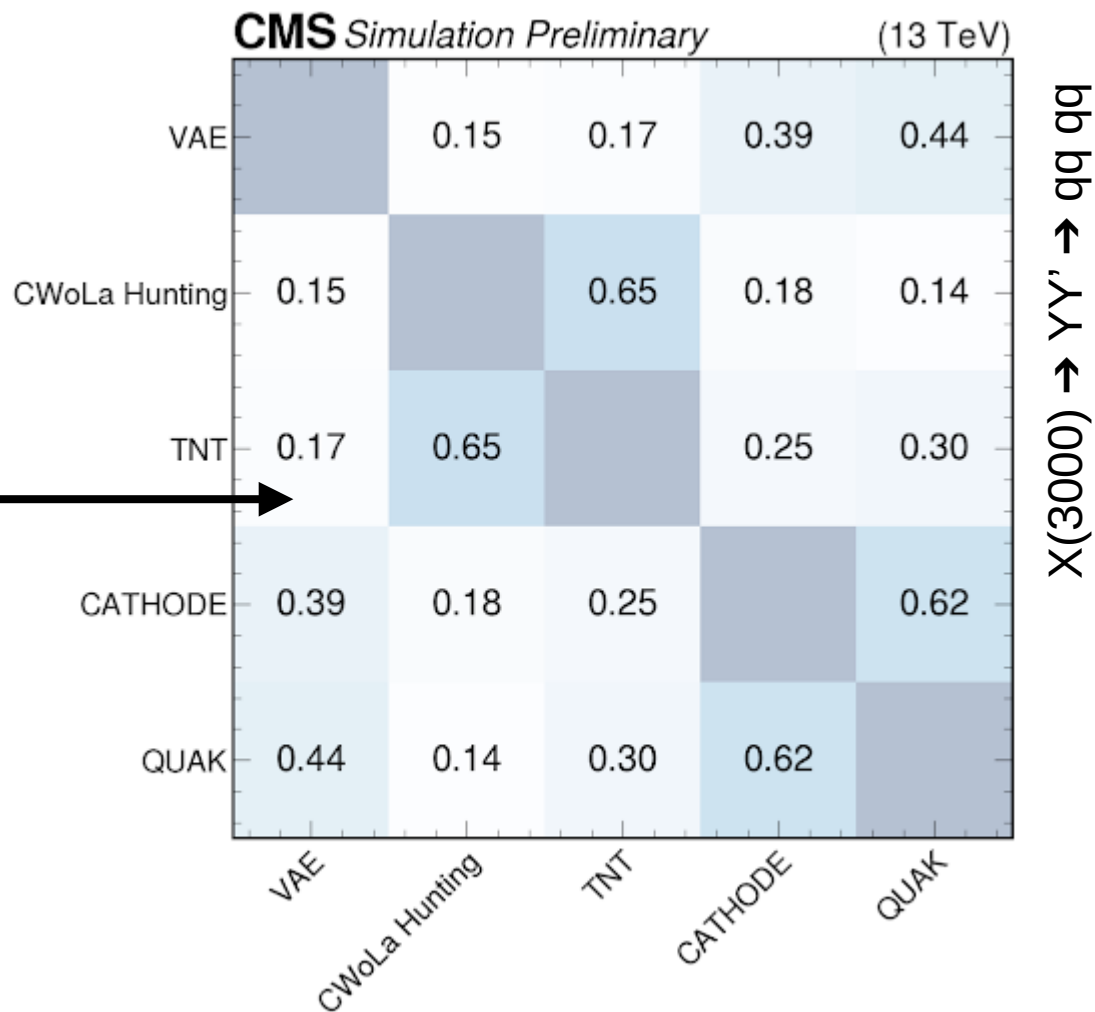
Signal:
 $X(3000) \rightarrow YY' \rightarrow qq\ qq$



Complementarity

Do all methods find the same events?

Check correlation between scores



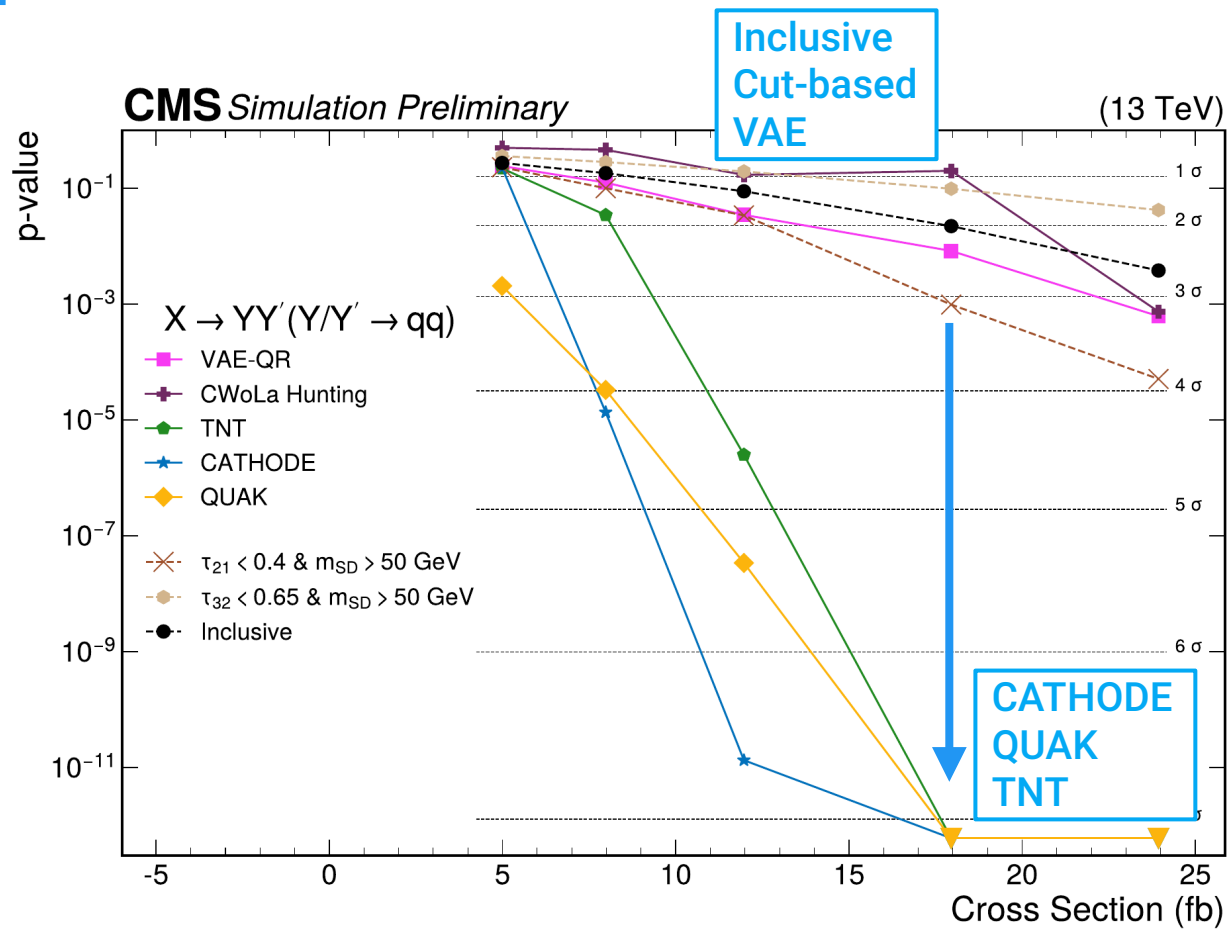
Small correlations

➔ Complementarity

Performance: 2 + 2

Testing performance on:
 $X(3000) \rightarrow YY' \rightarrow qq \, qq$

\simeq LHCO R&D dataset

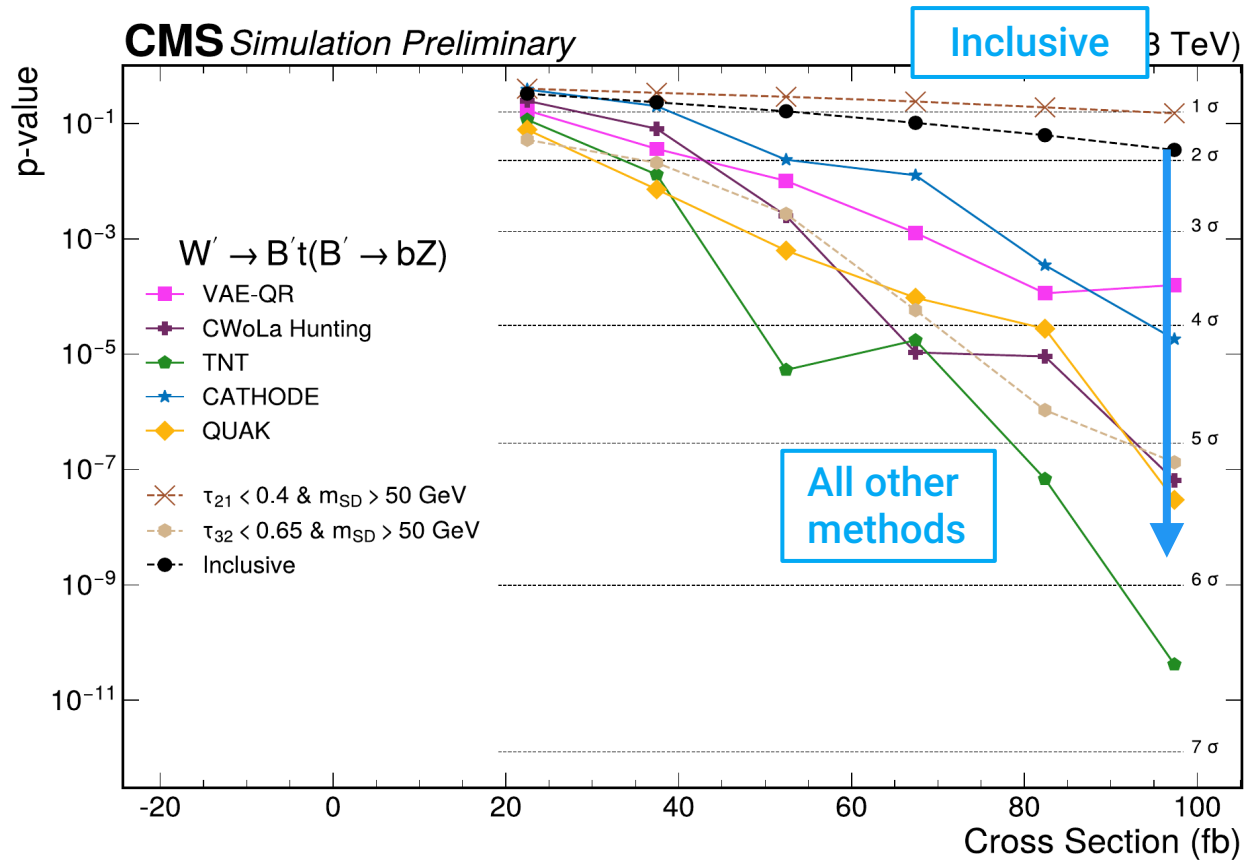


Performance: 3 + 3

Testing performance on:

$W' \rightarrow B't \rightarrow qqq qqq$

3 + 3



Performance: 3 + 3

Testing performance on:

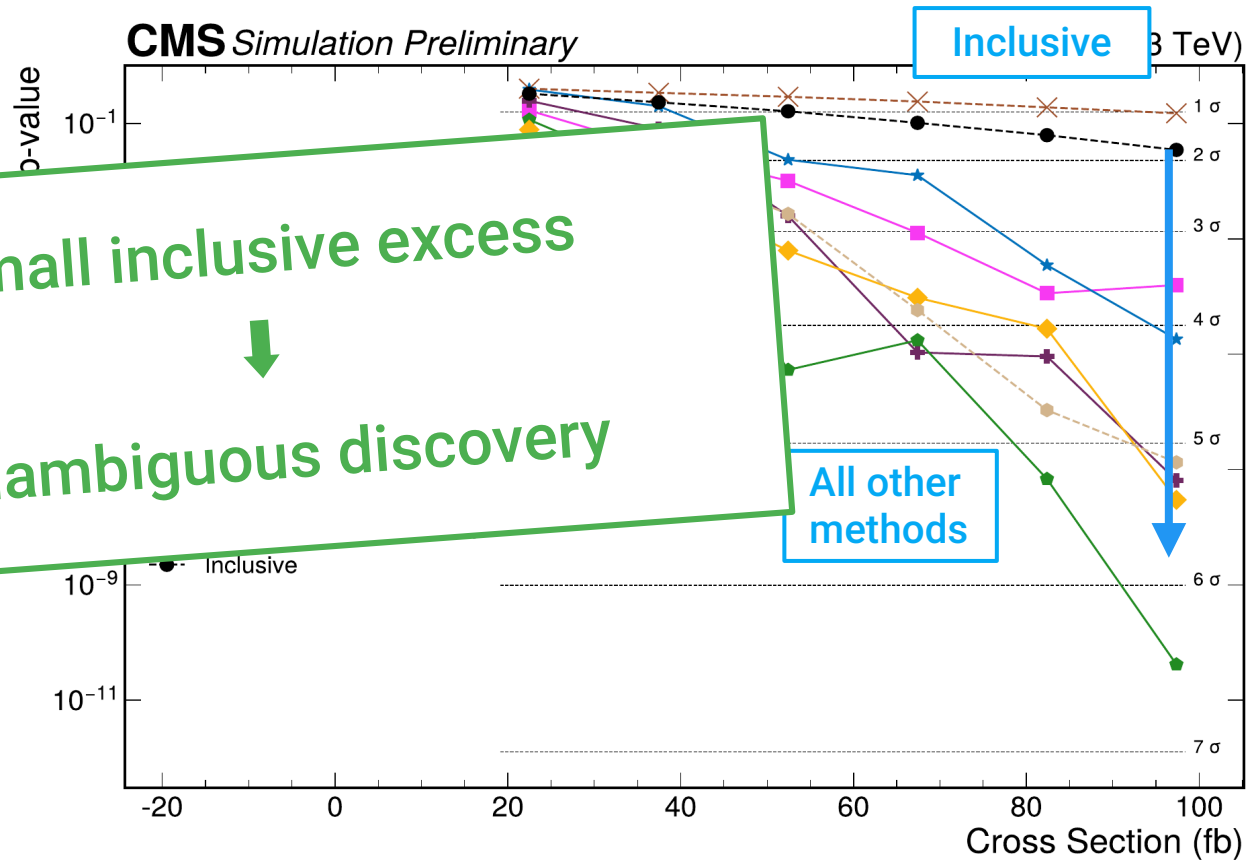
$W' \rightarrow B't \rightarrow qqq \ qqq$

3 +

Small inclusive excess



Unambiguous discovery

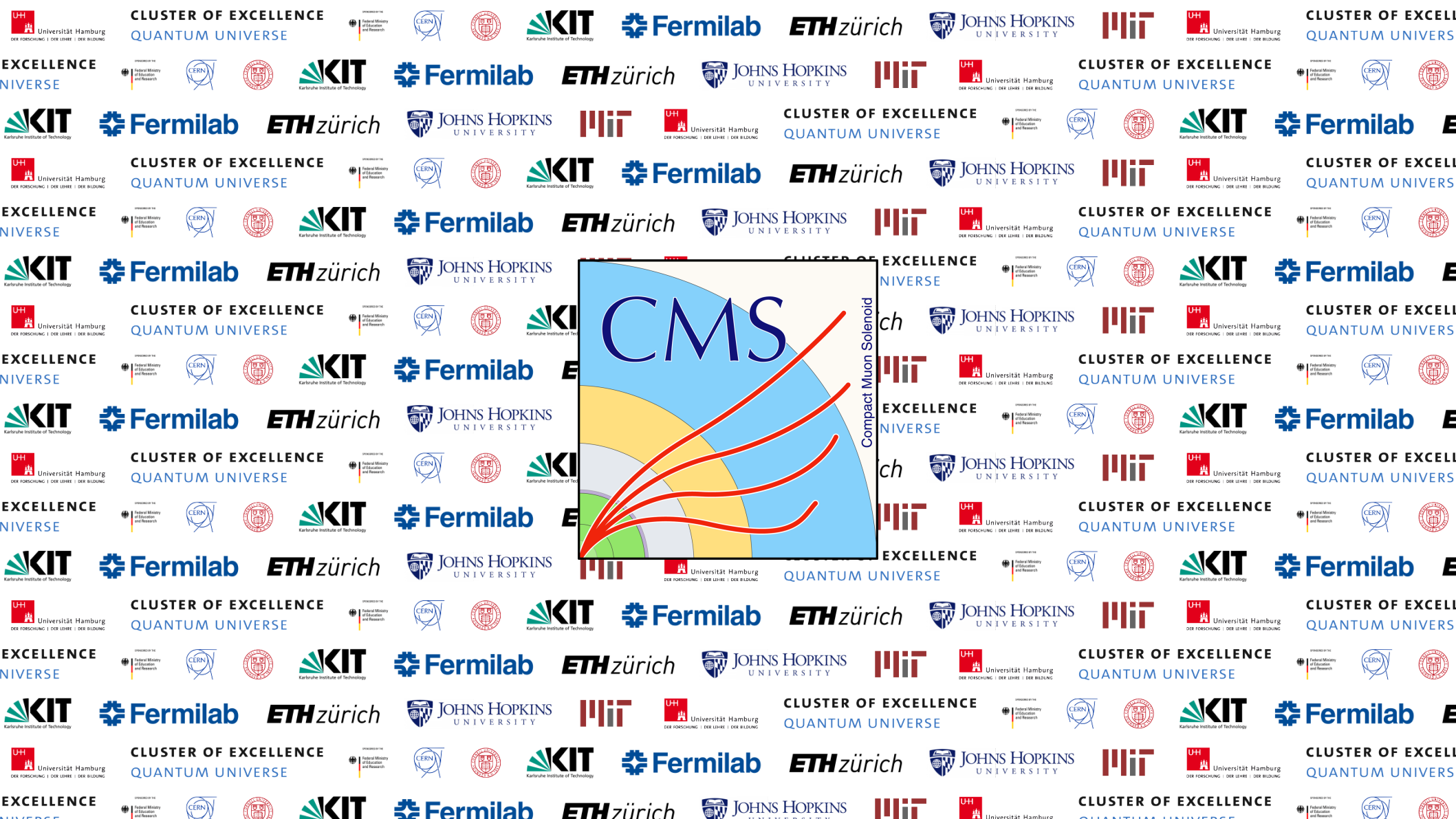


Summary

- CMS joining the anomaly detection party
- Looking for dijet resonances with **5** methods:
VAE, CWoLa, CATHODE, Tag N' Train, QUAK
- Promising performance in simulation
- No method to rule them all

Results to appear on CDS [\[link\]](#)

Finalizing analysis in data: stay tuned



Input Features

- **VAE:** p_T, η, ϕ of leading 100 particle flow constituents (per jet)
- **CWoLa, TNT:** $m_{SD}, \tau_{21}, \tau_{32}, \tau_{43}, n_{PF}, LSF_3, \text{b-tagging score}$ (per jet)
- **CATHODE:** $m_{SD1}, m_{SD1} - m_{SD2}, \tau_{41,1}, \tau_{41,2}$ (per event)
- **QUAK:** $m_{SD}, \tau_{21}, \tau_{32}, \tau_{43}, \sqrt{\tau_{21}/\tau_1}, M/p_T$ (for each jet, per event)

Complementarity

Do all methods find the same events?

Check correlation between scores (background)

Small bg correlations

