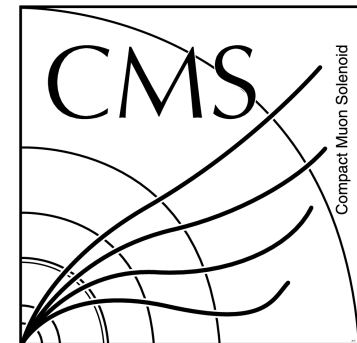


Dealing with Large backgrounds




Phil Harris
(MIT)



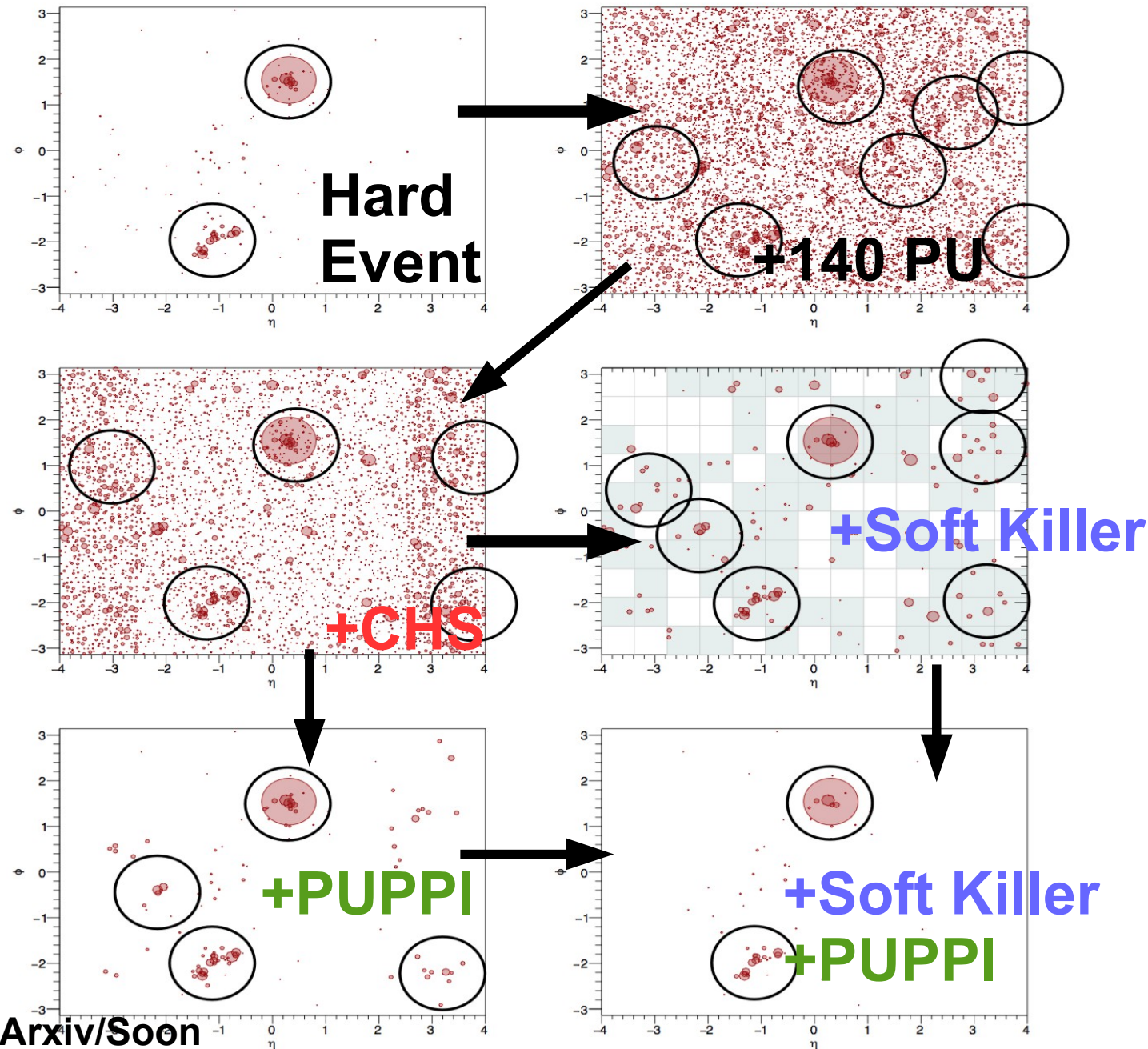
Overview

- Contending with UE makes jet properties difficult
 - Typical energy densities have $\rho > 200 \text{ GeV}$
 - That translates to $> 100 \text{ GeV/AK4 jet area}$
 - Inside the jet UE still does not look like a jet
 - Jets contain most of their energy in $\Delta R < 0.1$ (7 GeV)
 - Means there are ways to highlight jet specifics
 - However we cannot necessarily remove everything
- Dealing with large backgrounds in pp
 - In pp collisions there are a few more handles
 - However the challenges and precision are different
 - In this talk we draw an analogy with a problem in pp

What are the features of UE?

- UE is the result of many low energy processes
 - Consists of low energy particles
 - Spread roughly uniformly over the event
 - Not part of the hard core of a jet
 - For PP : not part of the primary vertex
 - This has led to a number of core algorithms
 - Soft Killer : cut particles below a p_T threshold
 - Constituent/p subtraction : remove avg energy/particle
 - PUPPI : compute likelihood from a jet fragmentation
 - CHS : remove charged particles not from PV
 - Note: at this stage pp prefers particle based sub.
- 

How does this look?



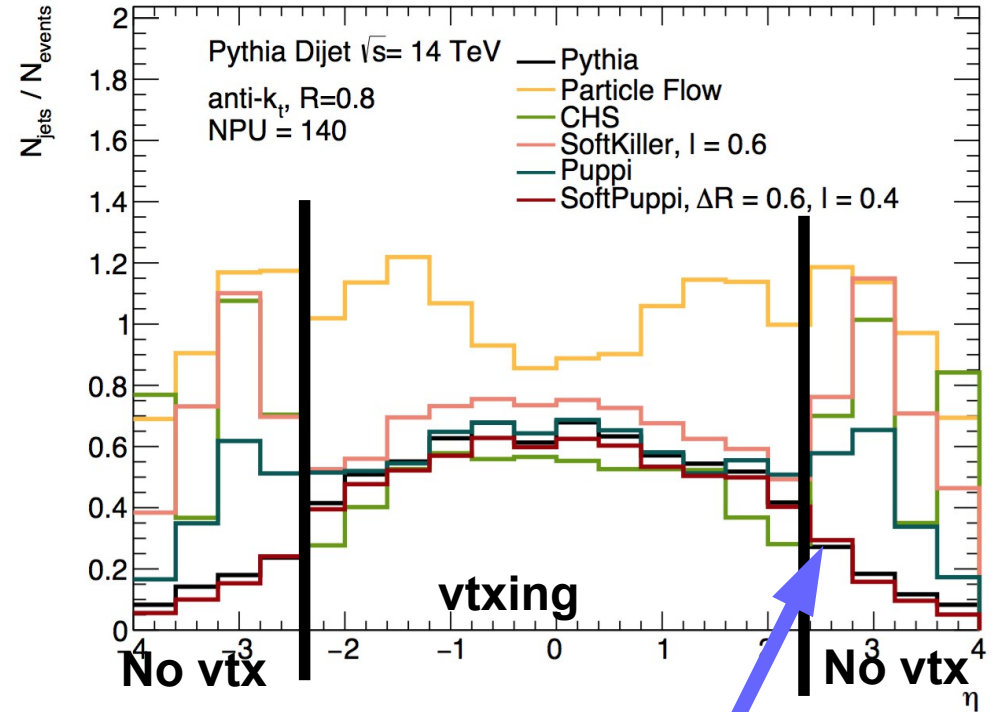
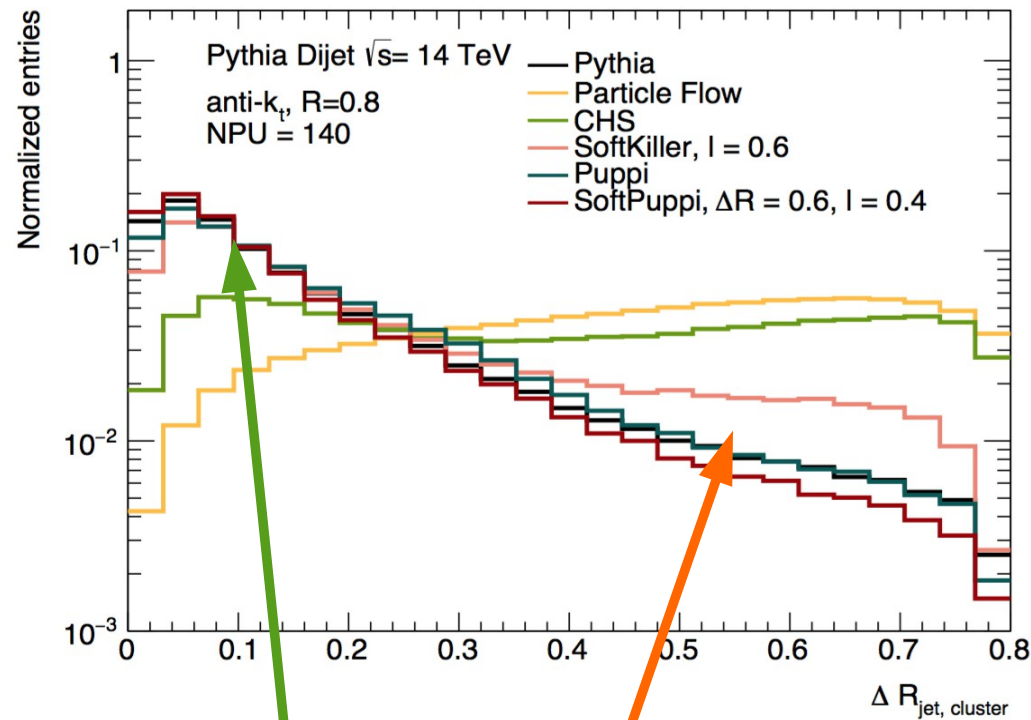
Effectively :

Fragmentation
uses radial
information

p_T uses hard
scatter

Combination of
the two brings
most robust
performance

How does this affect the shape?



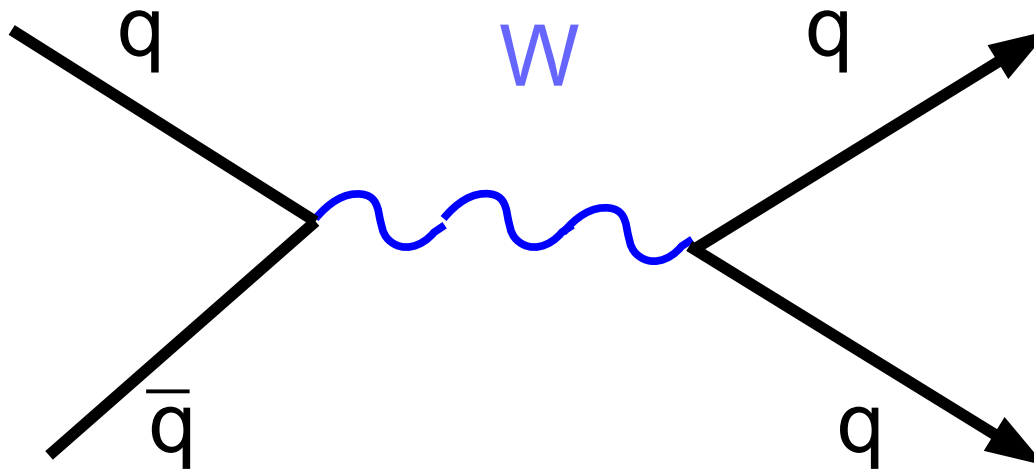
- Using radial information biases jets narrow
- Using p_T information biases jets wide
- Using p_T information removes fake jets

Drawing the analogy with HI

- Despite the success of pp PU(UE) removal algos
 - PP physics still has to contend with degradation
 - Not all UE removal can be performed perfectly
 - Forward region is still an active area of study
 - However impact is much smaller than in HI
- Background subtraction is still a critical issue :
 - This critical issue occurs in jet physics
 - Tools we use to resolve this issue are relevant to HI
 - Lets consider a relevant example with a lot of bkg
 - This is a measurement I would love to do in the future

The Problem

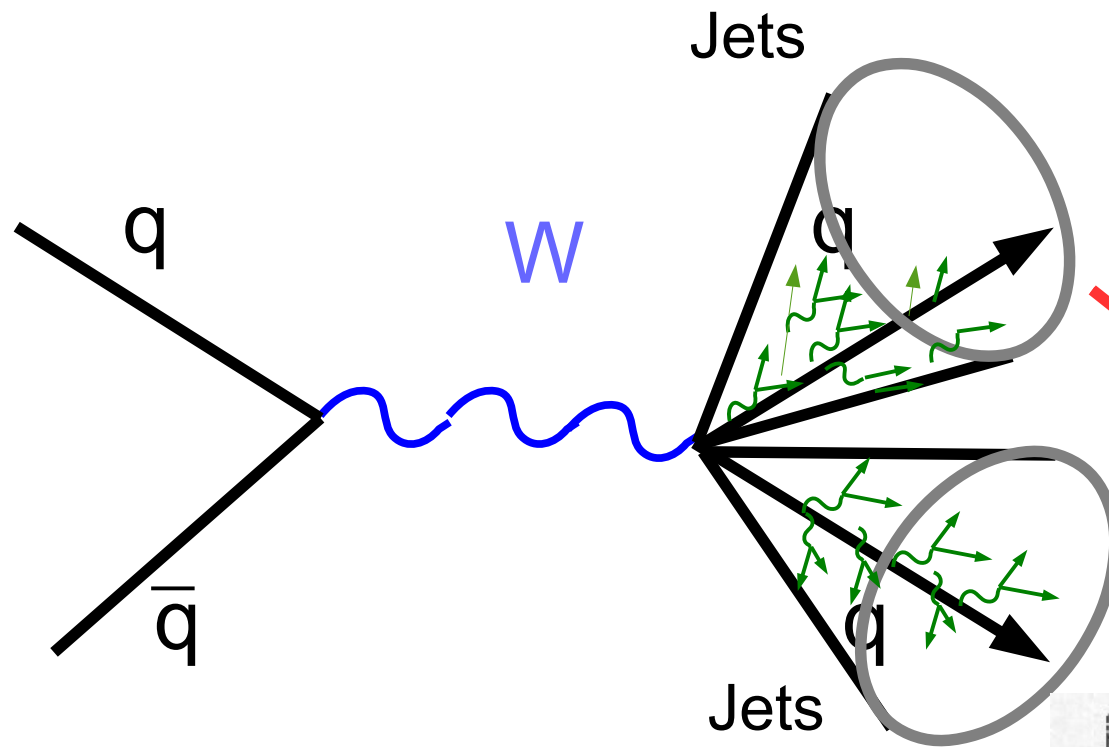
Find:



A W boson decaying to quarks

In an experiment

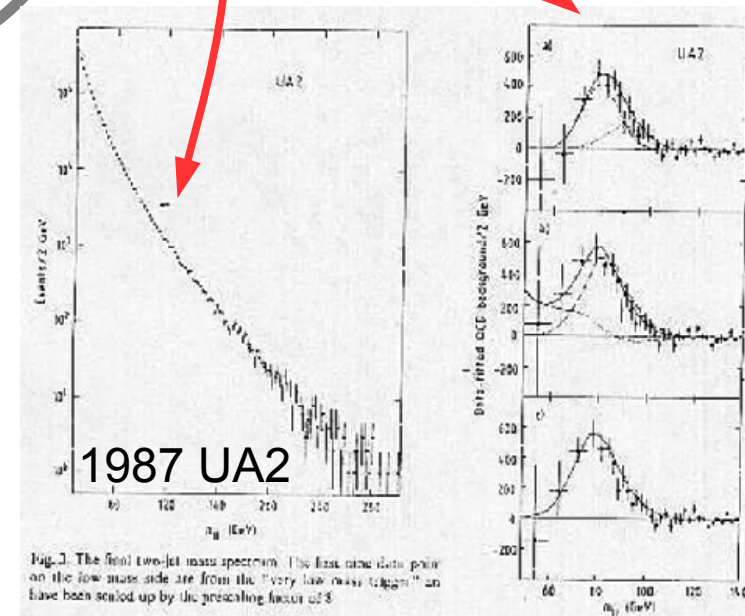
Find:



Resolved as a tiny bump on a huge background

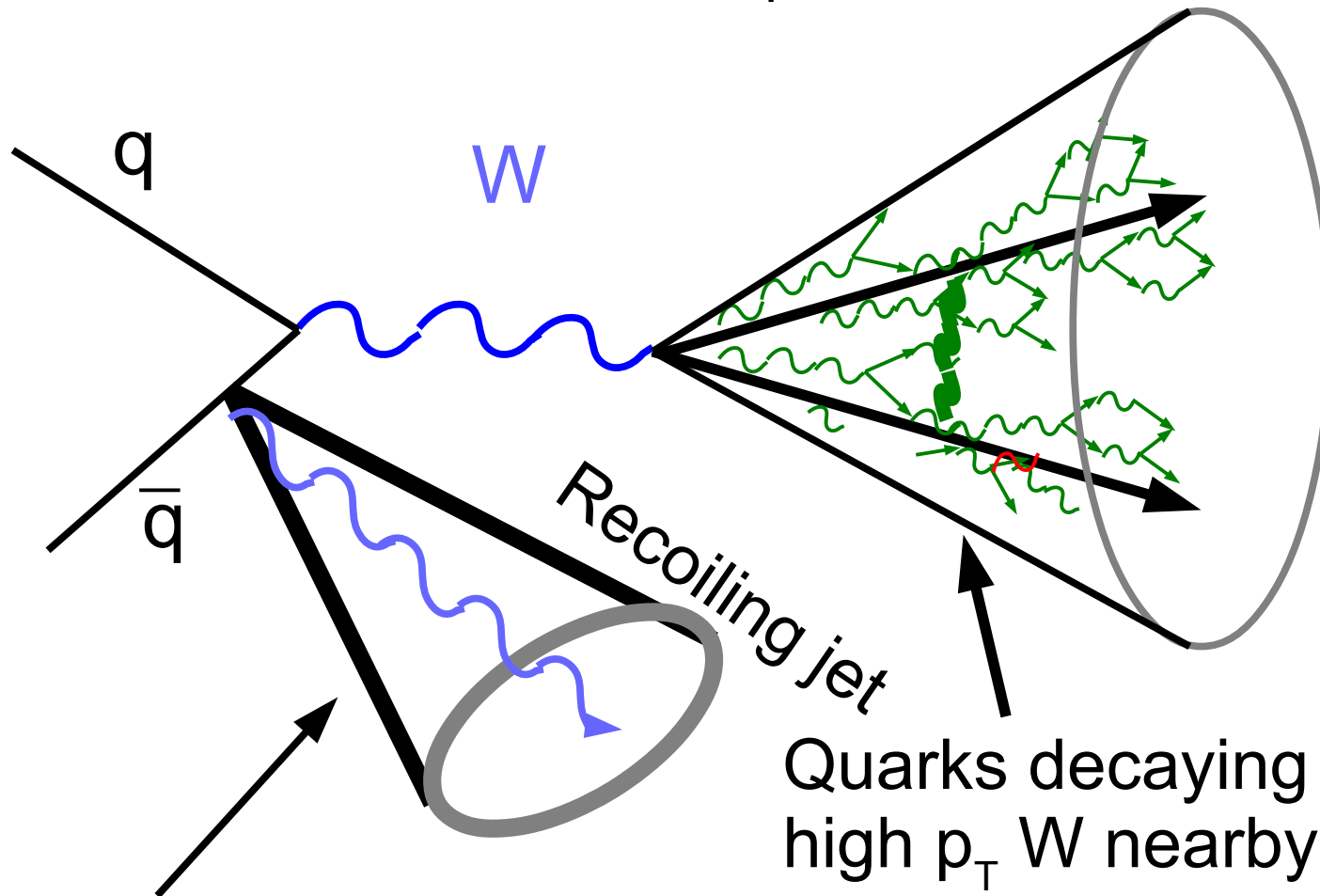
A W boson decaying to quarks
A low p_T we resolve them as jets

Was last observed 30 years
Result was not very convincing
Have not seen this since



Whats the first goal?

Require W to be a high p_T

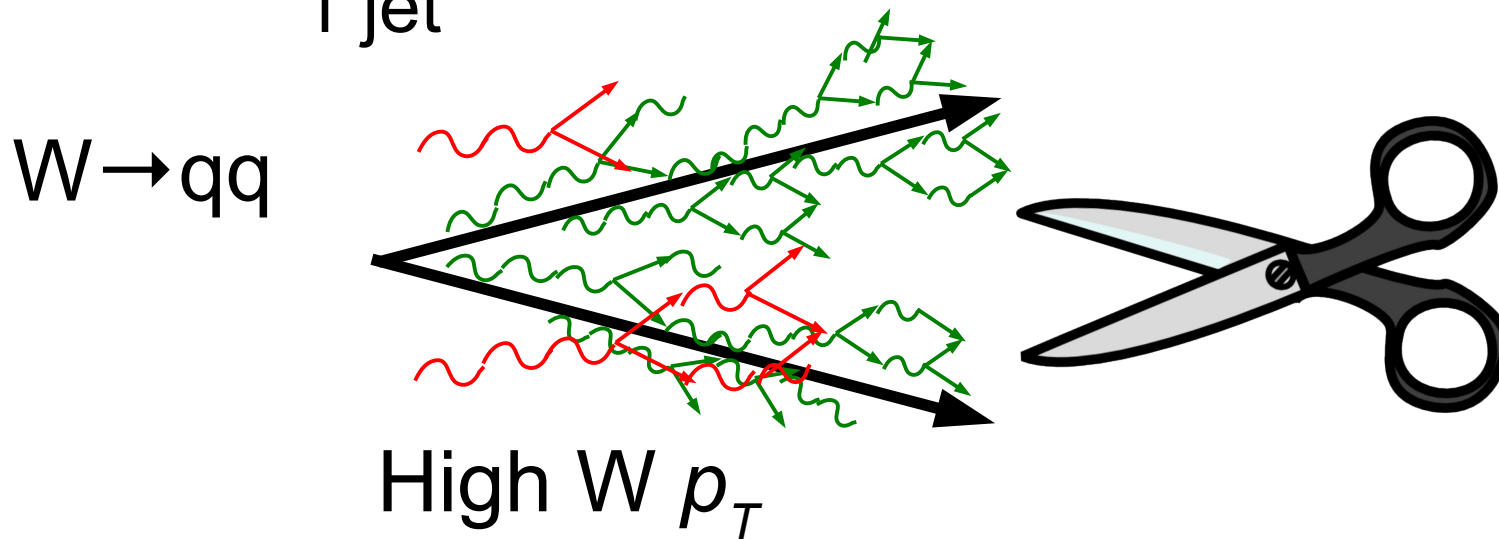


Recoiling jet off W gives the W high p_T

Quarks decaying from high p_T W nearby each other
Resolve this as a single jet

Jet grooming

- In order get a clean mass peak:
 - Adopt an approach that cleans excess radiation in a jet

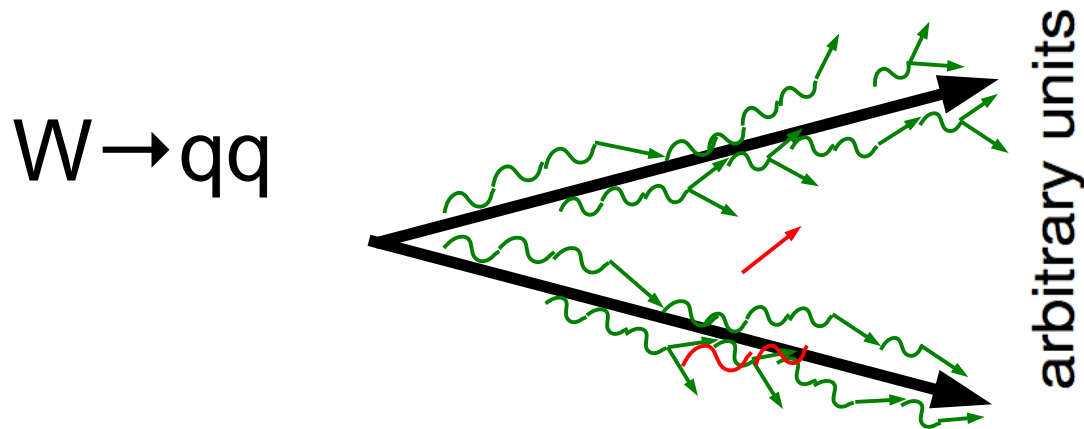


Pileup/QCD radiation



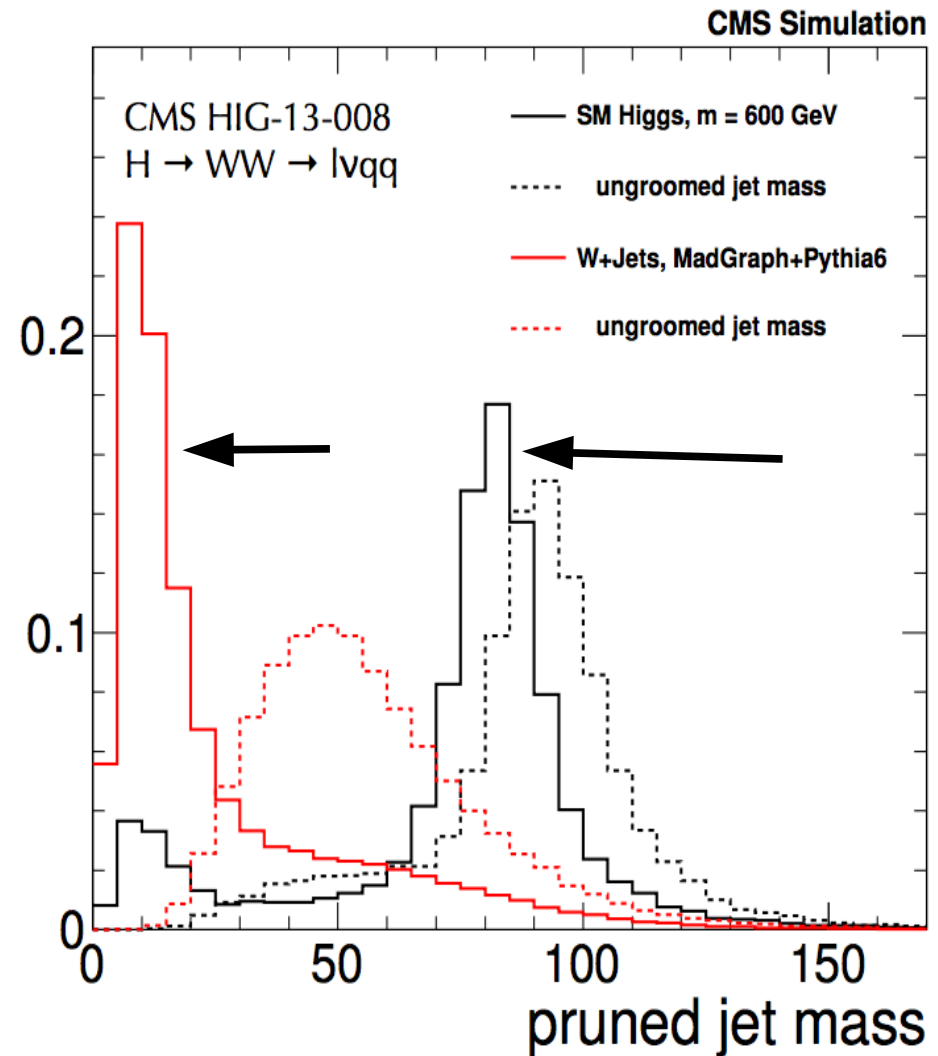
Jet grooming

- In order get a clean mass peak:
 - Adopt an approach that cleans excess radiation in a jet



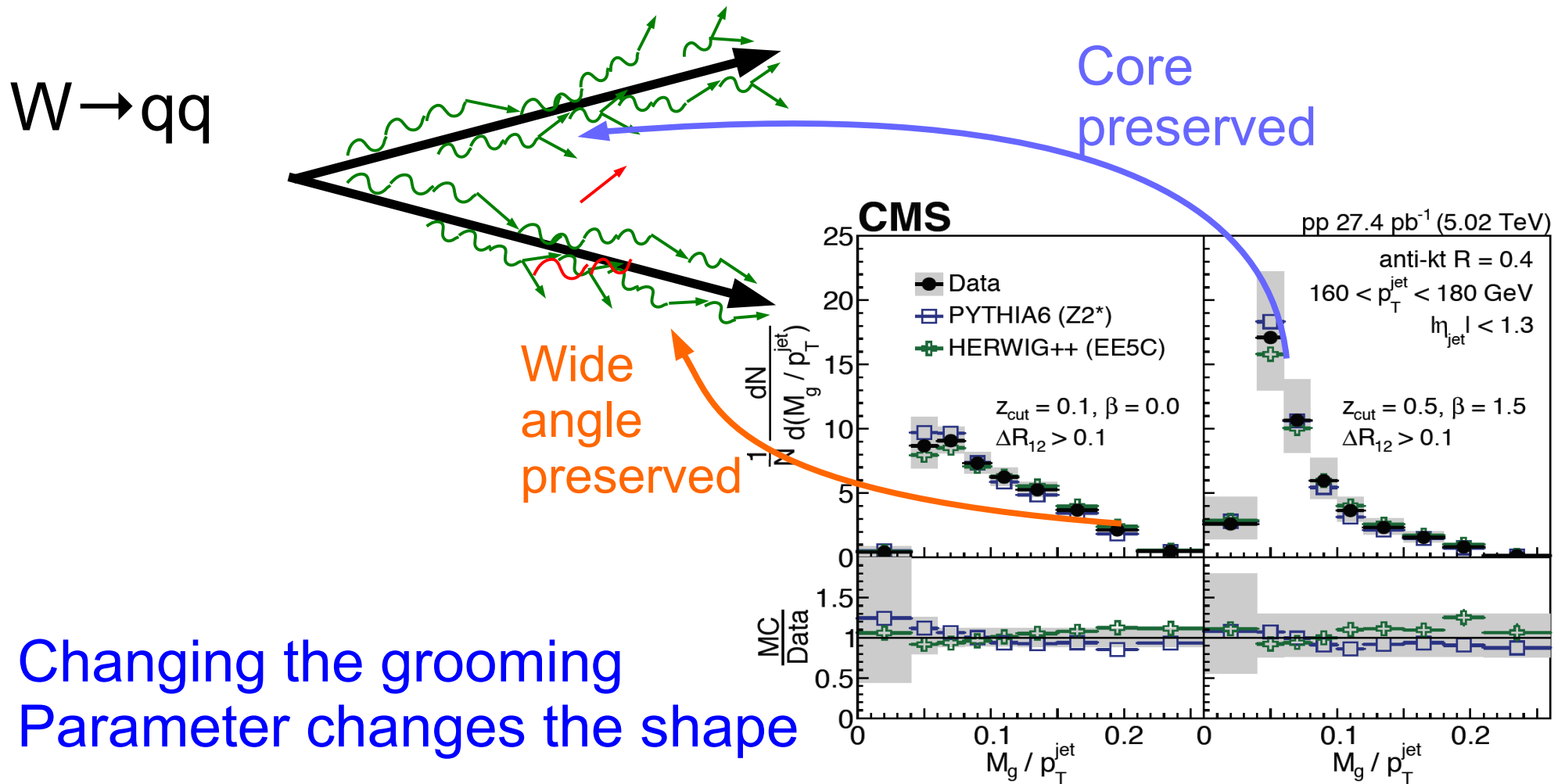
Jet grooming improves the resolution of jets

Pileup/QCD radiation



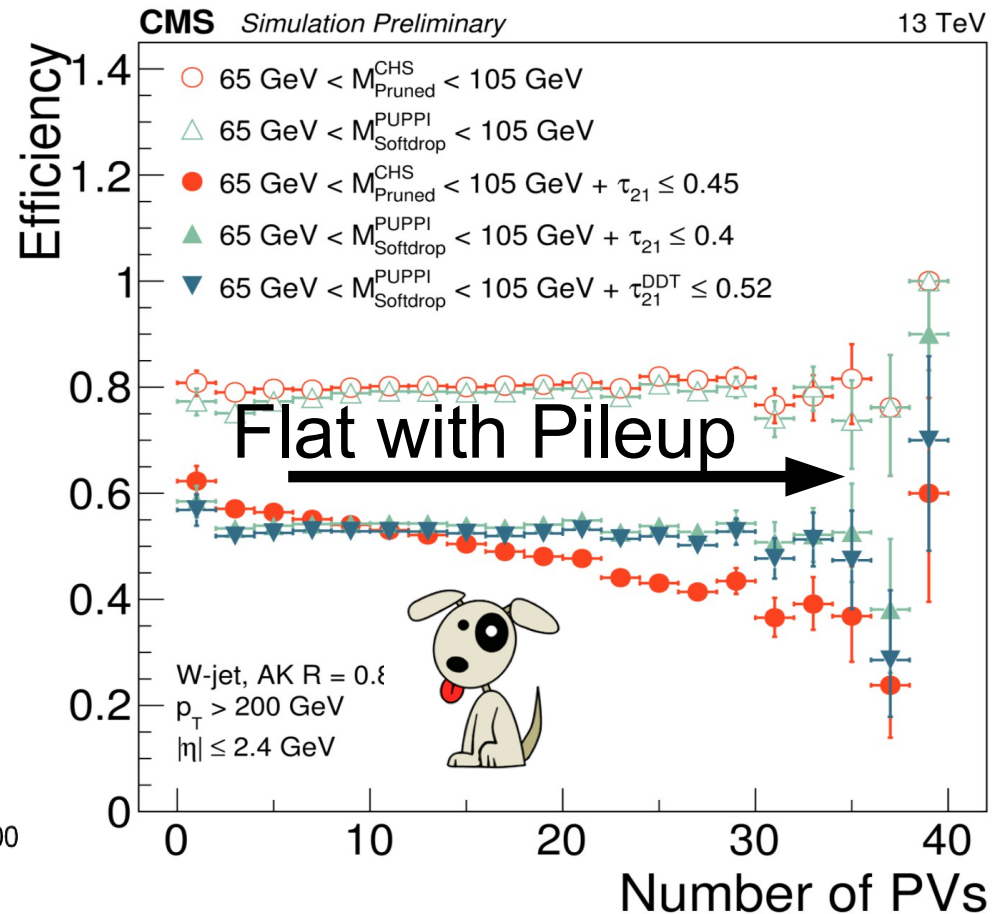
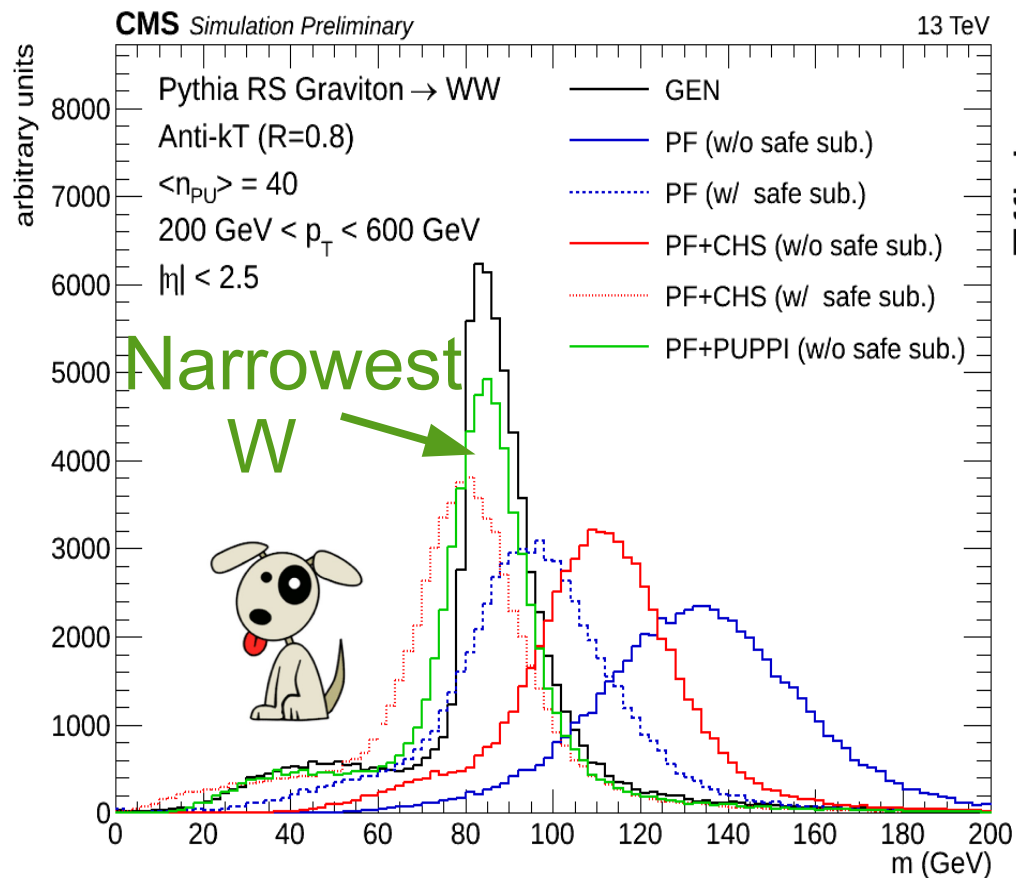
Jet grooming

- In order get a clean mass peak:
 - Adopt an approach that cleans excess radiation in a jet



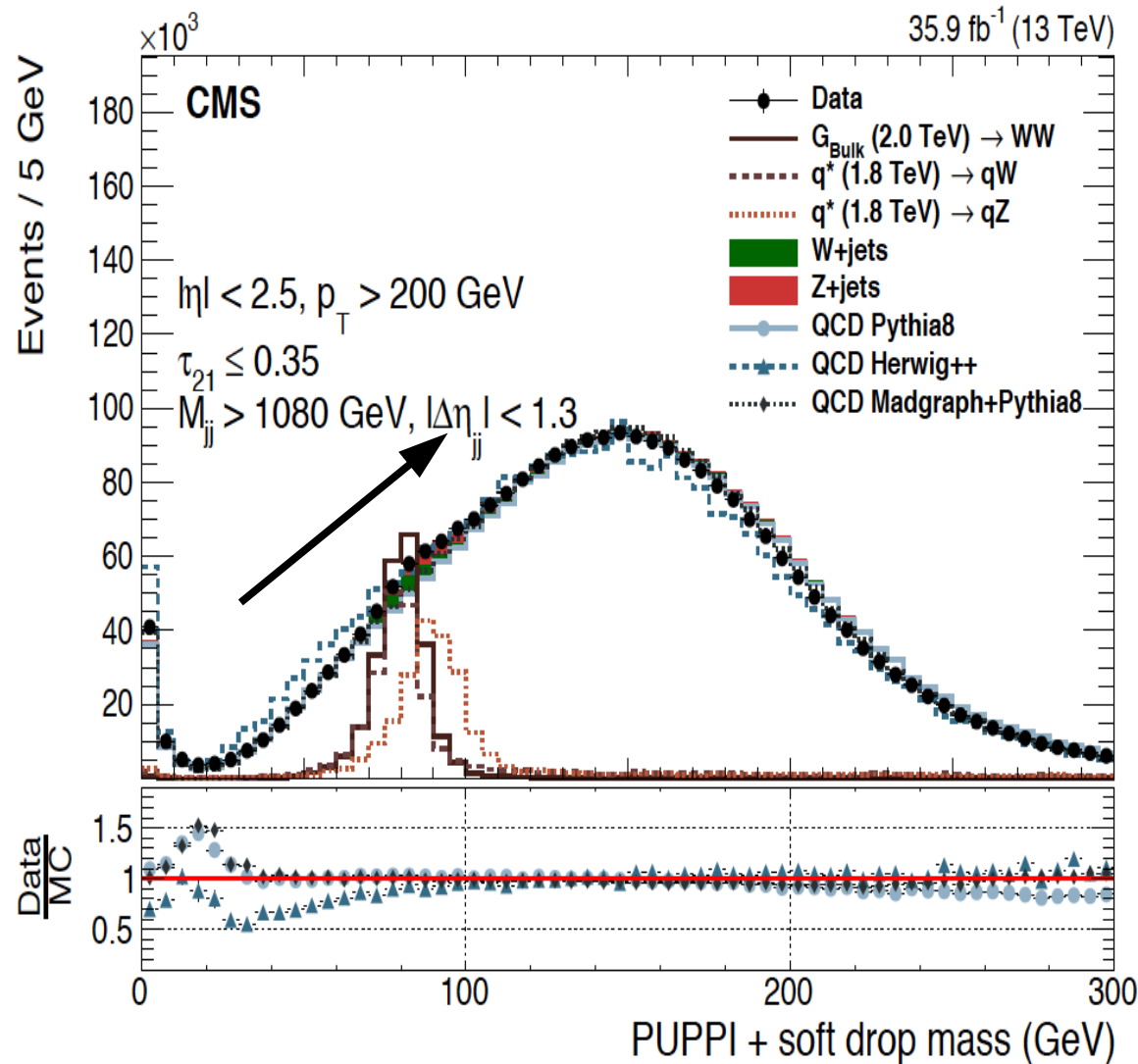
Additional Useful point in PP

- To reduce impact of pileup on substructure
 - Utilize the PUPPI algorithm + Softdrop $\beta=0$



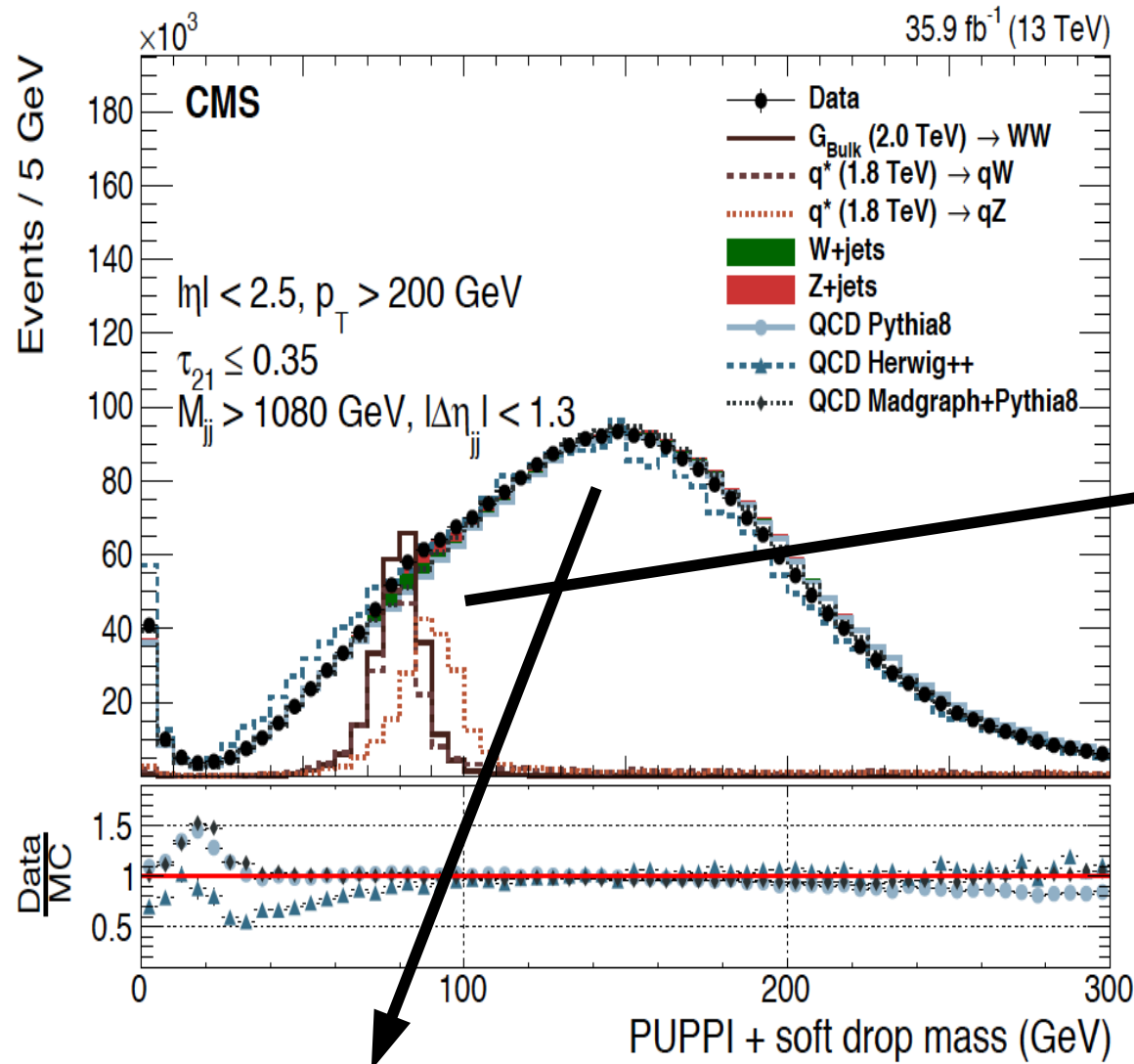
PUPPI gives best performance under PU conditions

Finding the W/Z on a slope



Background is peaking after the bump
 helps to improve reduce background at W

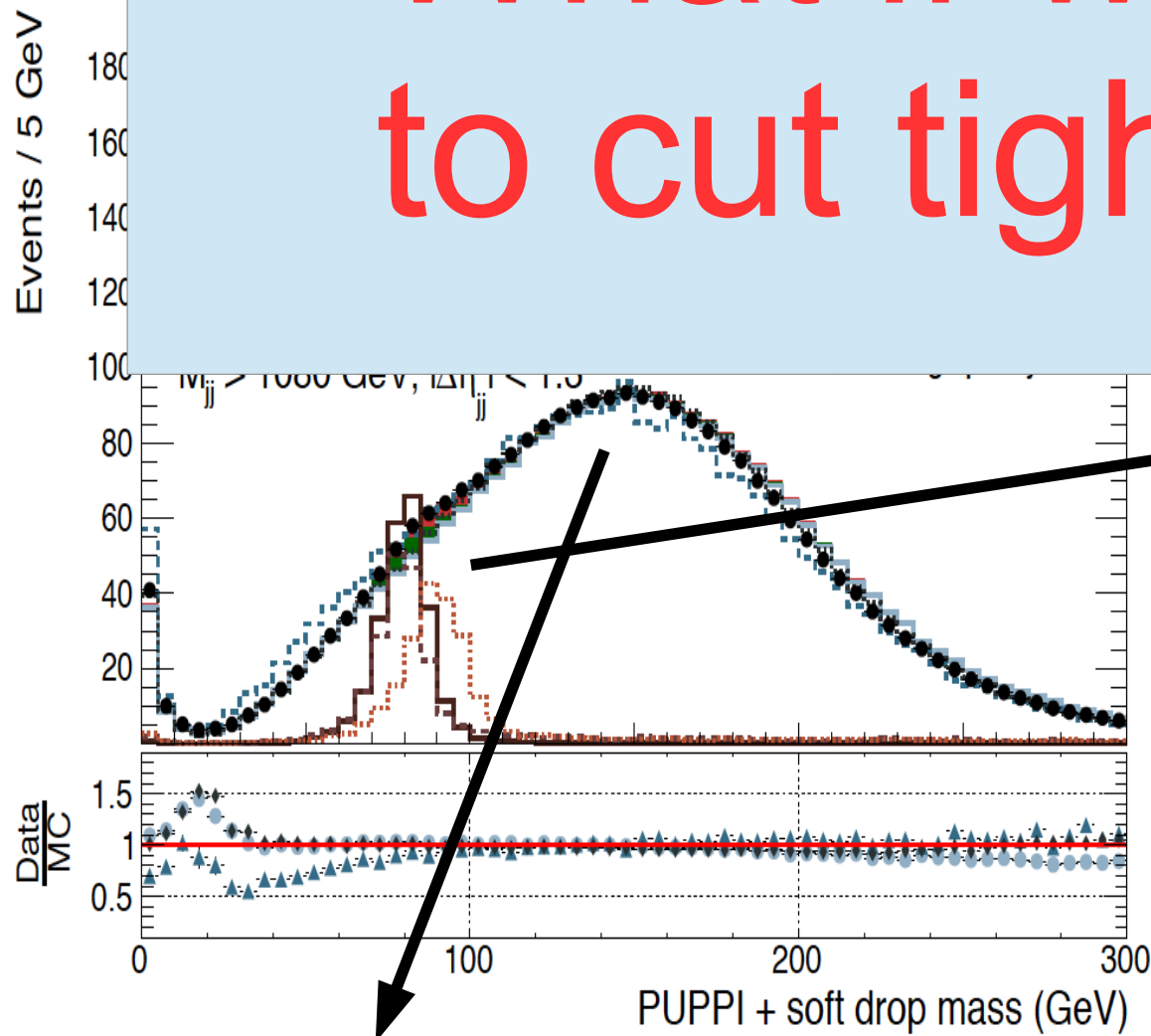
Finding the W/Z on a slope



How does the upward slope impact the search for a resonance?

Background is peaking after the bump
 helps to improve reduce background at W

What if we want to cut tighter?



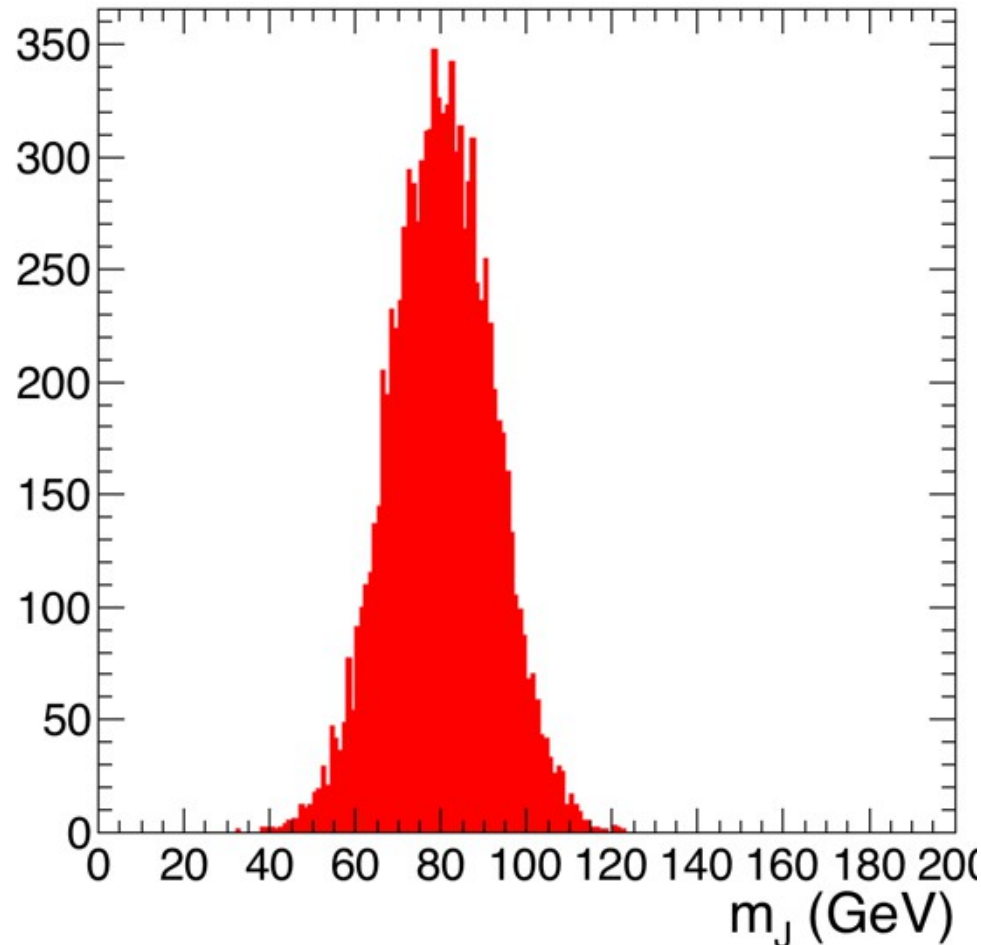
upward slope impact
the search for a
resonance?

Background is peaking after the bump
helps to improve reduce background at W

A Quick Game

Lets play a game

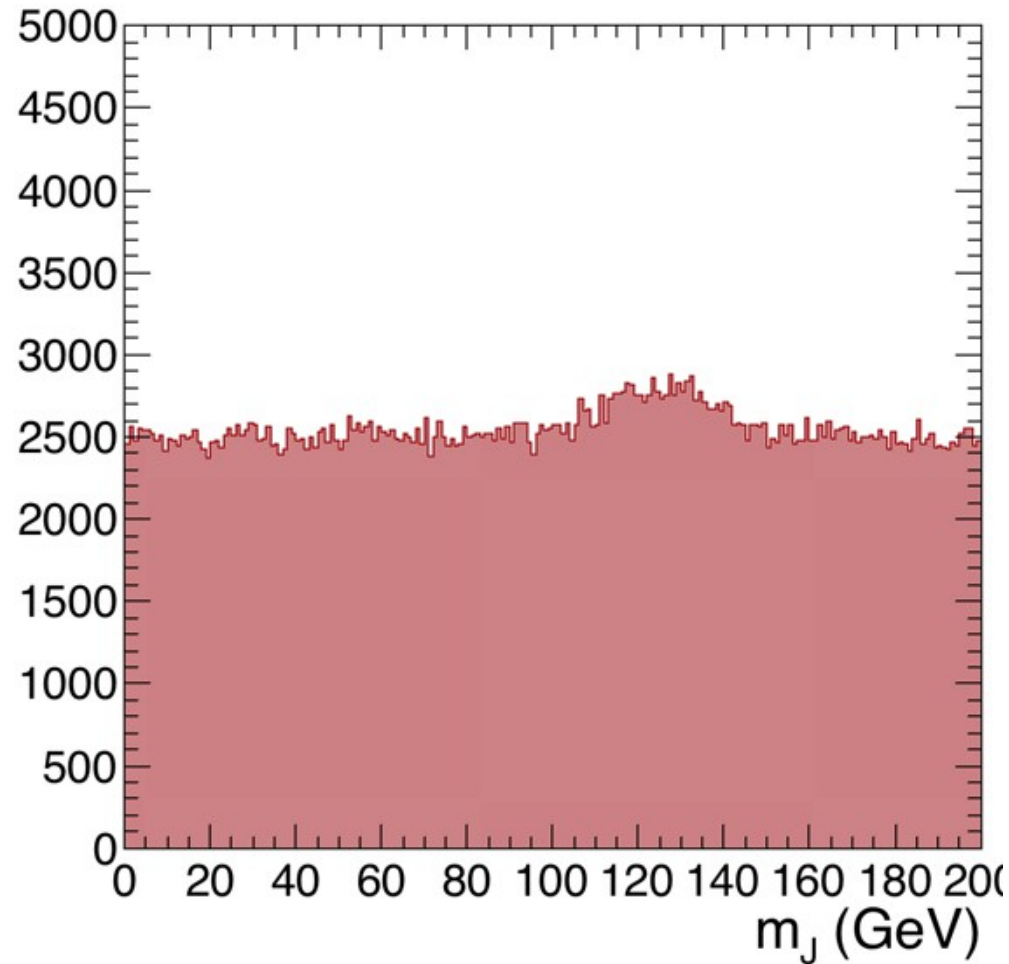
- Lets try to inject a signal likes this



Into a background 50 times larger than this
This signal looks very much like a W peak

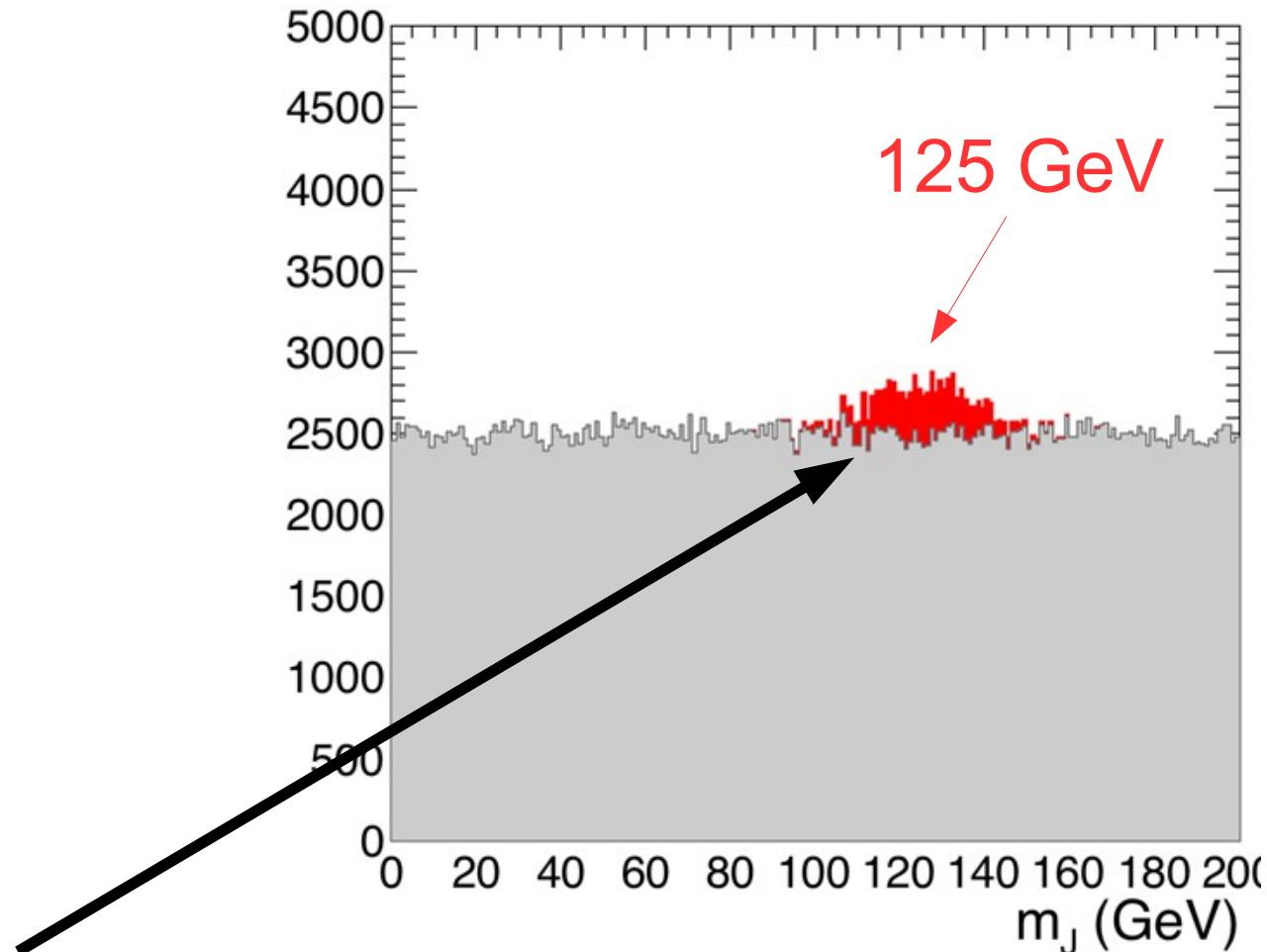
Find the signal

- Can you see the signal?



Find the signal

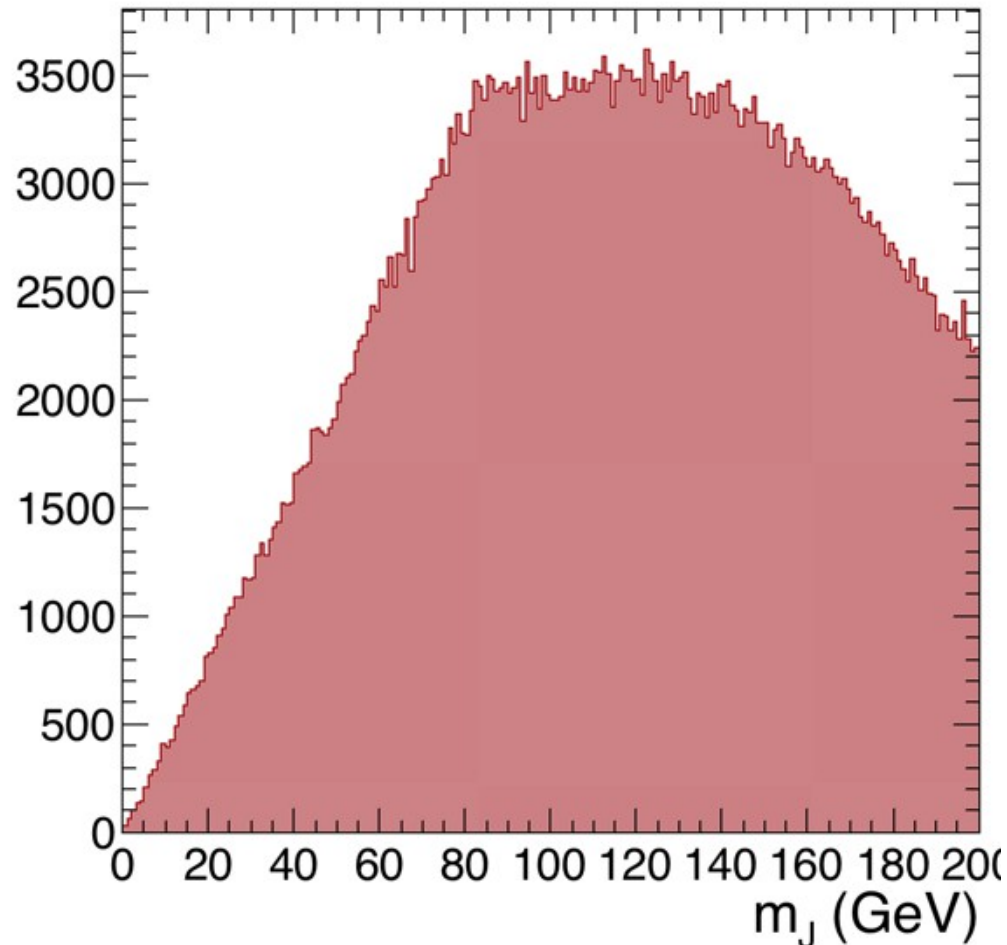
- Can you see the signal?



Here is the signal at 125 GeV on a flat background
That was an easy one!

Find the signal

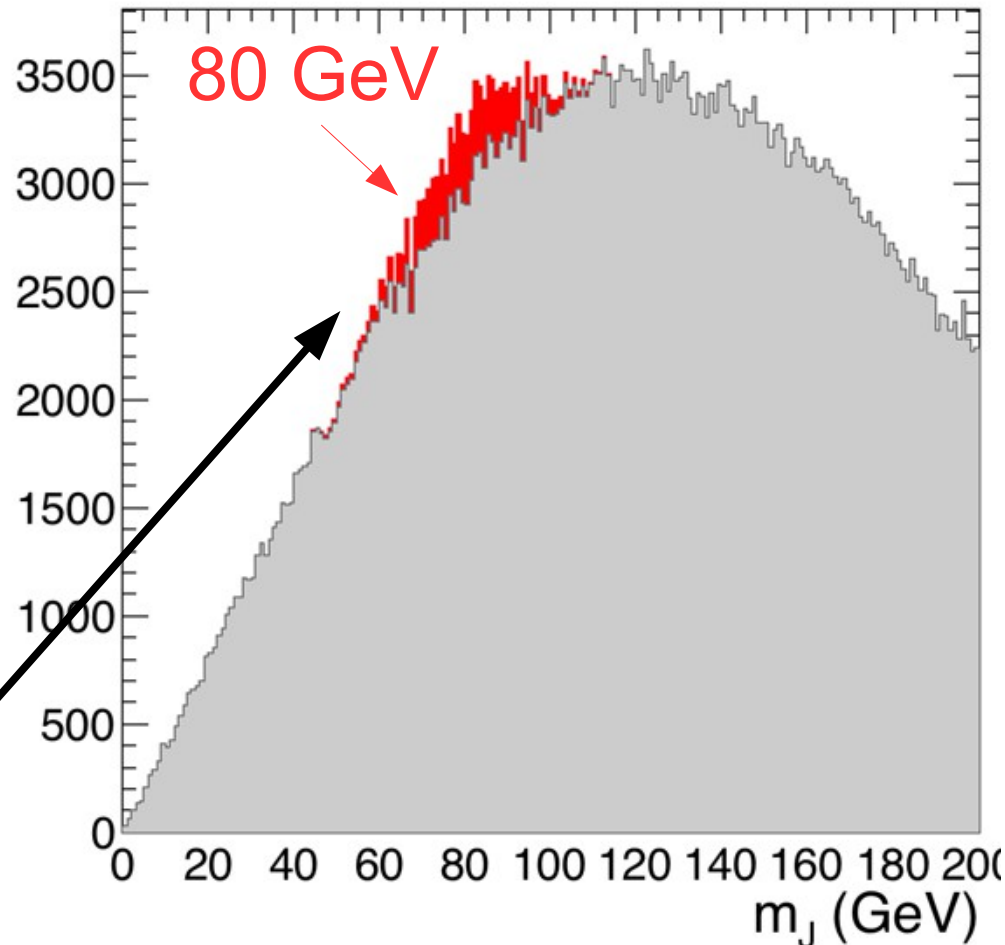
- Can you see the signal?



Here is the signal at 125 GeV on a flat background
Now the background peaks just after the signal

Find the signal

- Can you see the signal?

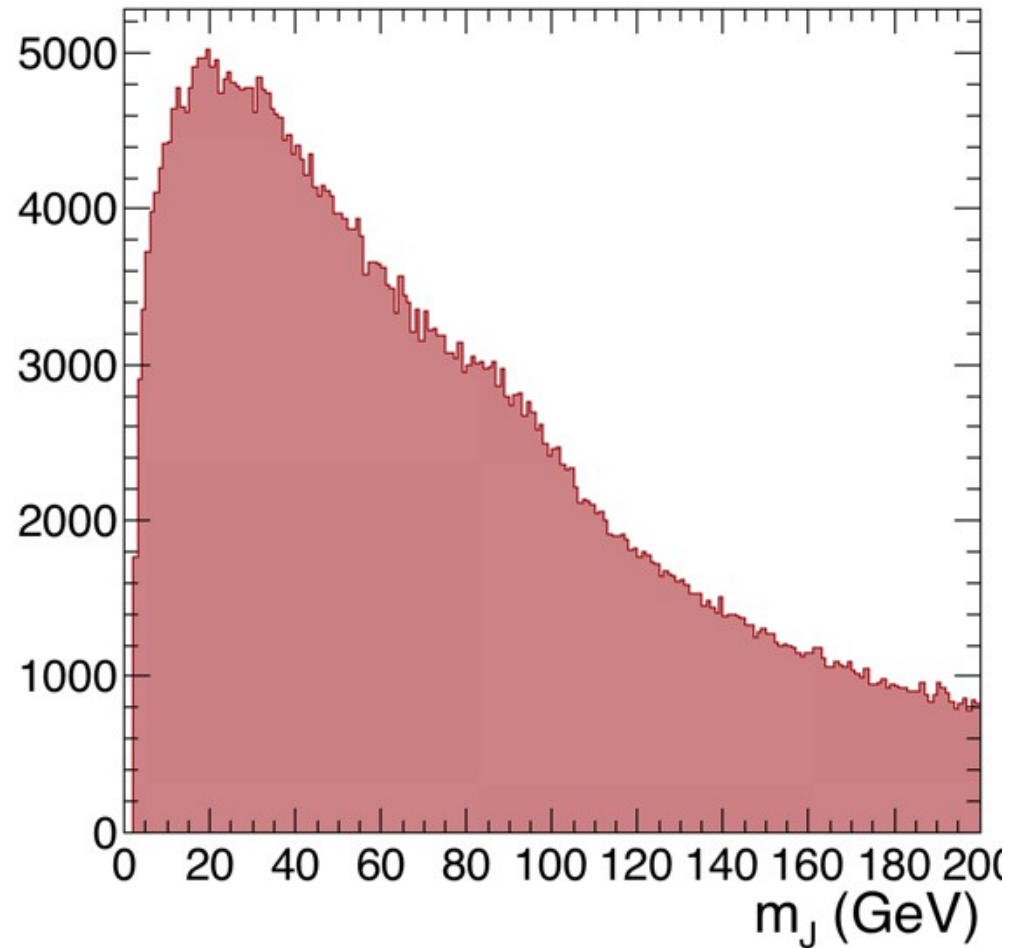


Here it is. This one is a bit harder.

Note exact same amount of background and signal

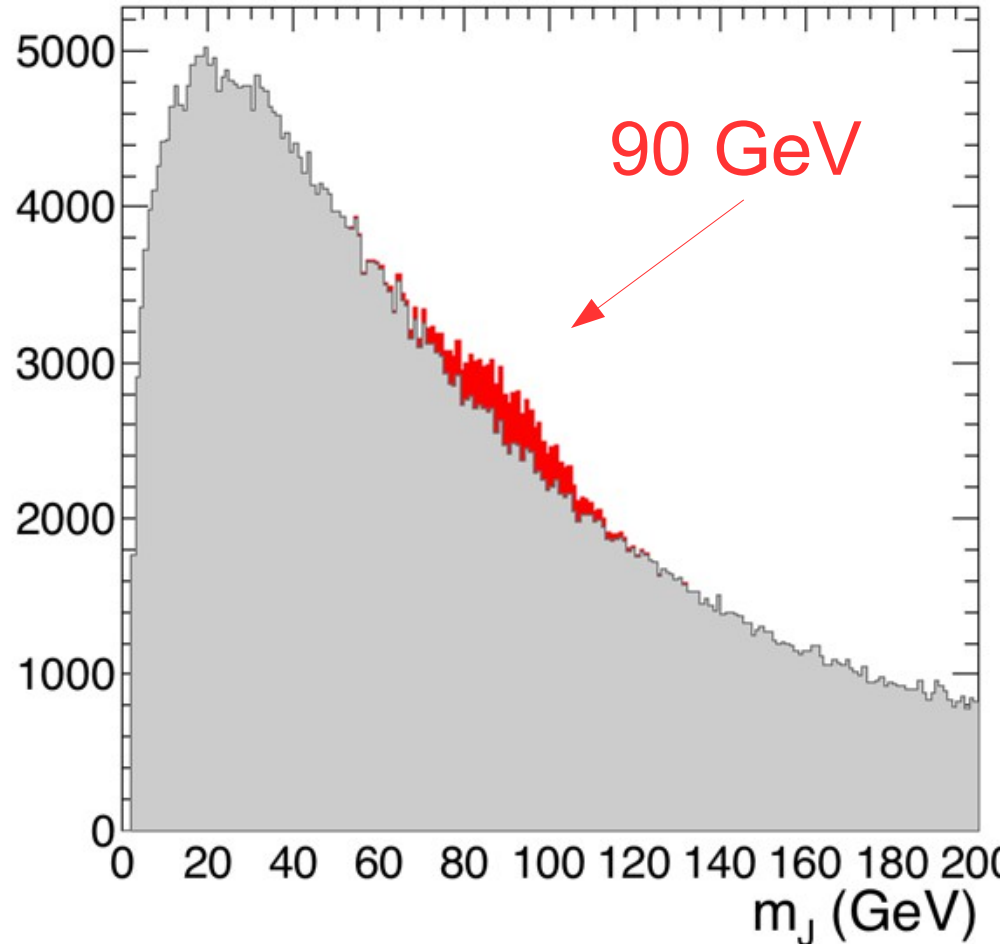
How about now?

- Can you see the signal?



How about now?

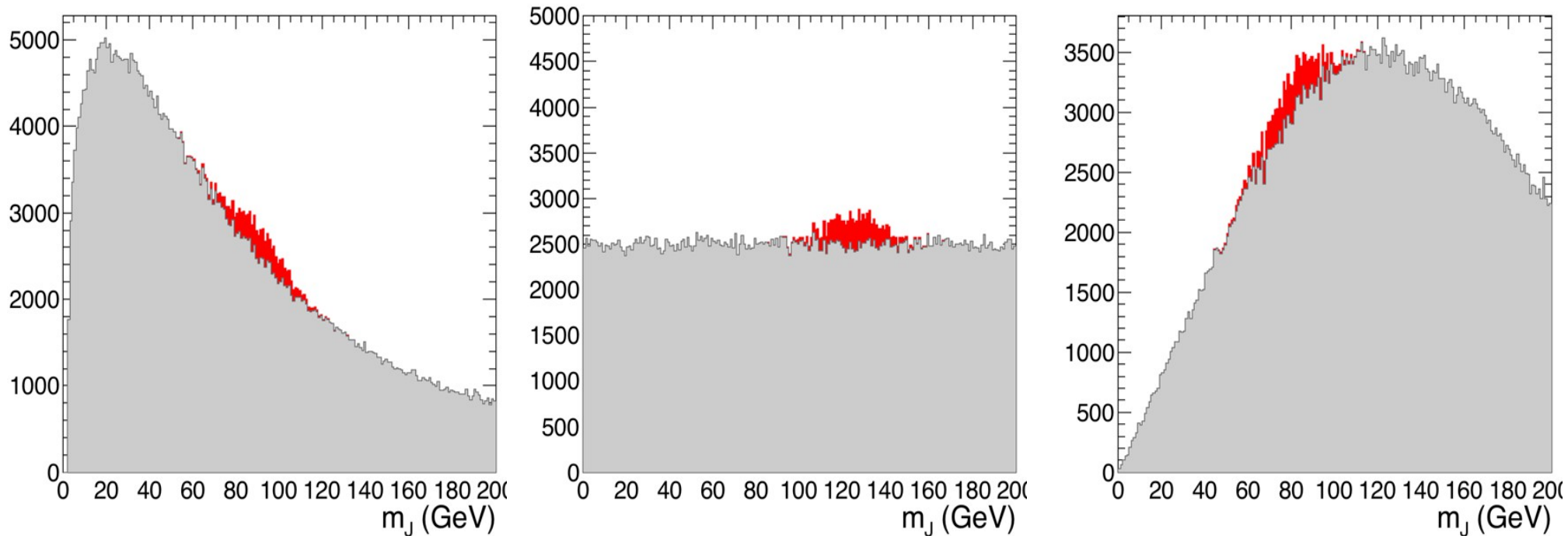
- Can you see the signal?



Background

Note exact same amount of background and signal

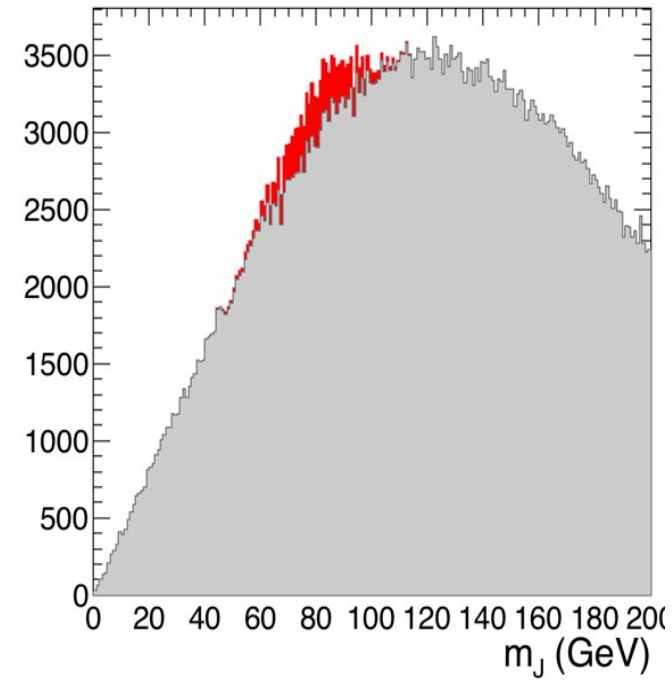
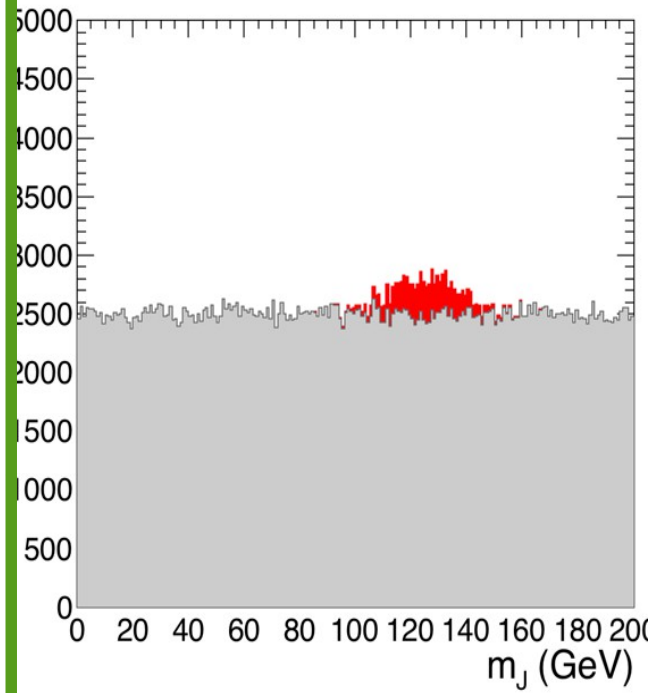
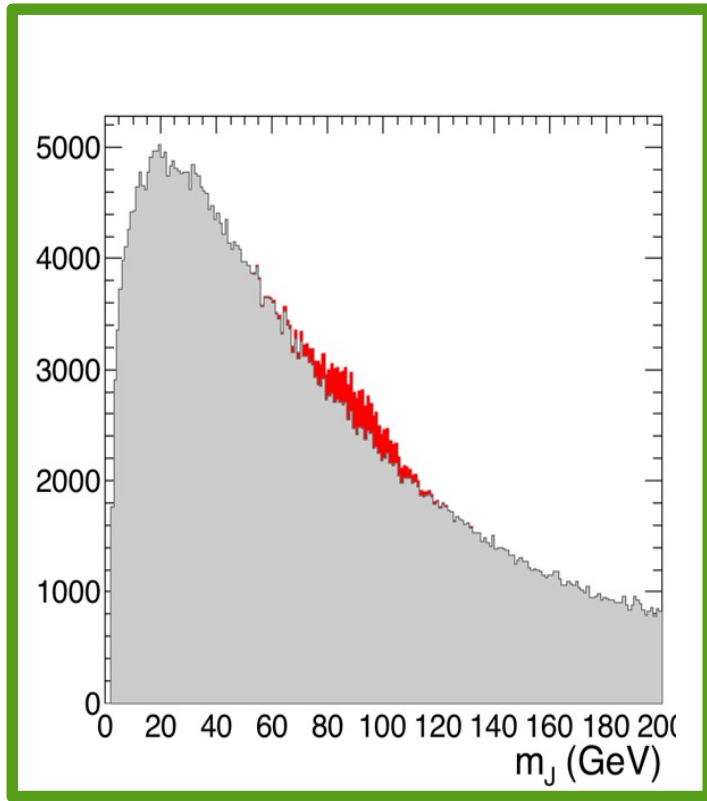
Whats best background?



All variables yield roughly the same performance
Exactly the same background yield
Exactly the same signal yield

What if we could tune background shape to what we want?
What would be our best choice of shape?

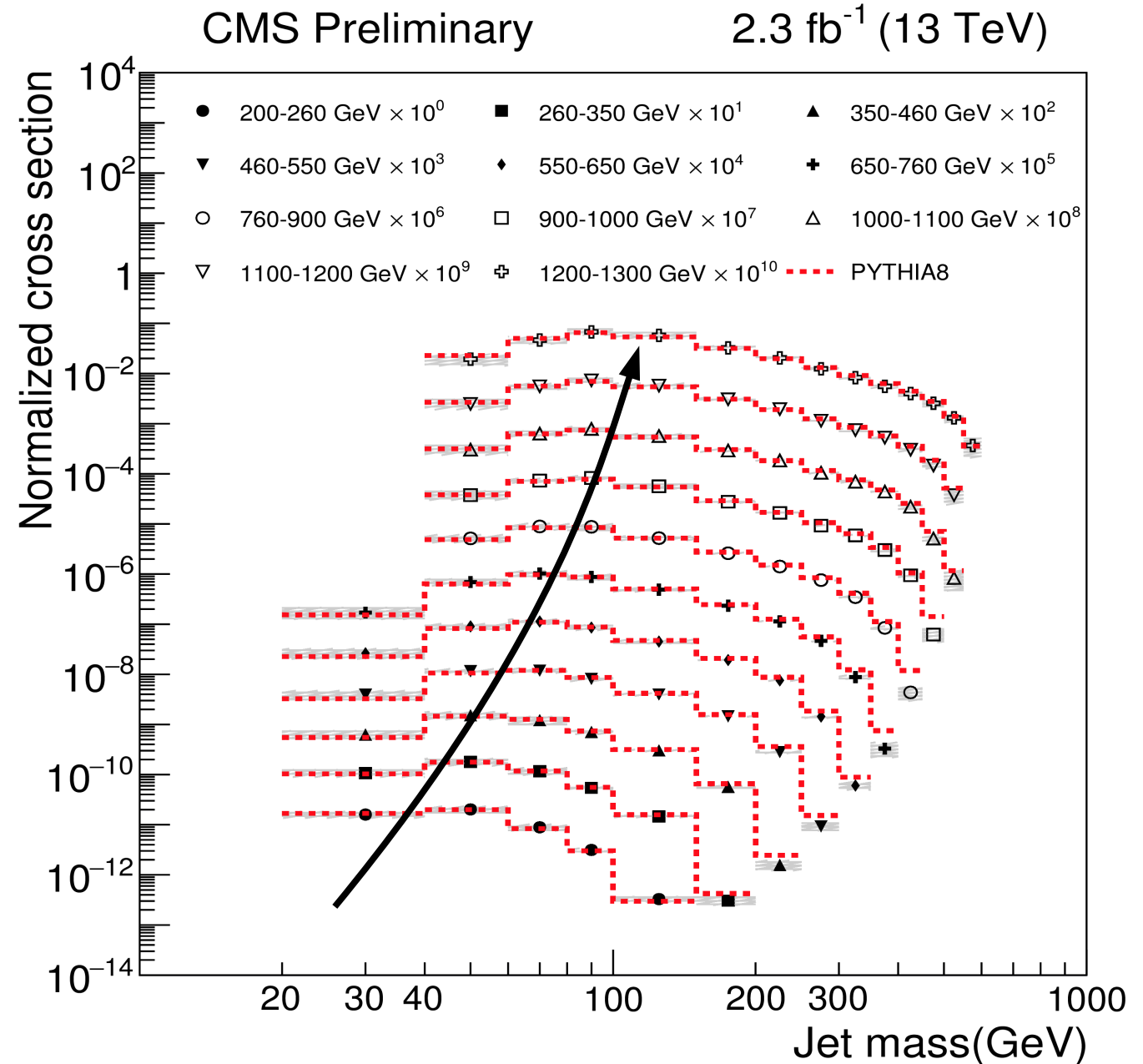
Whats best background?



This one

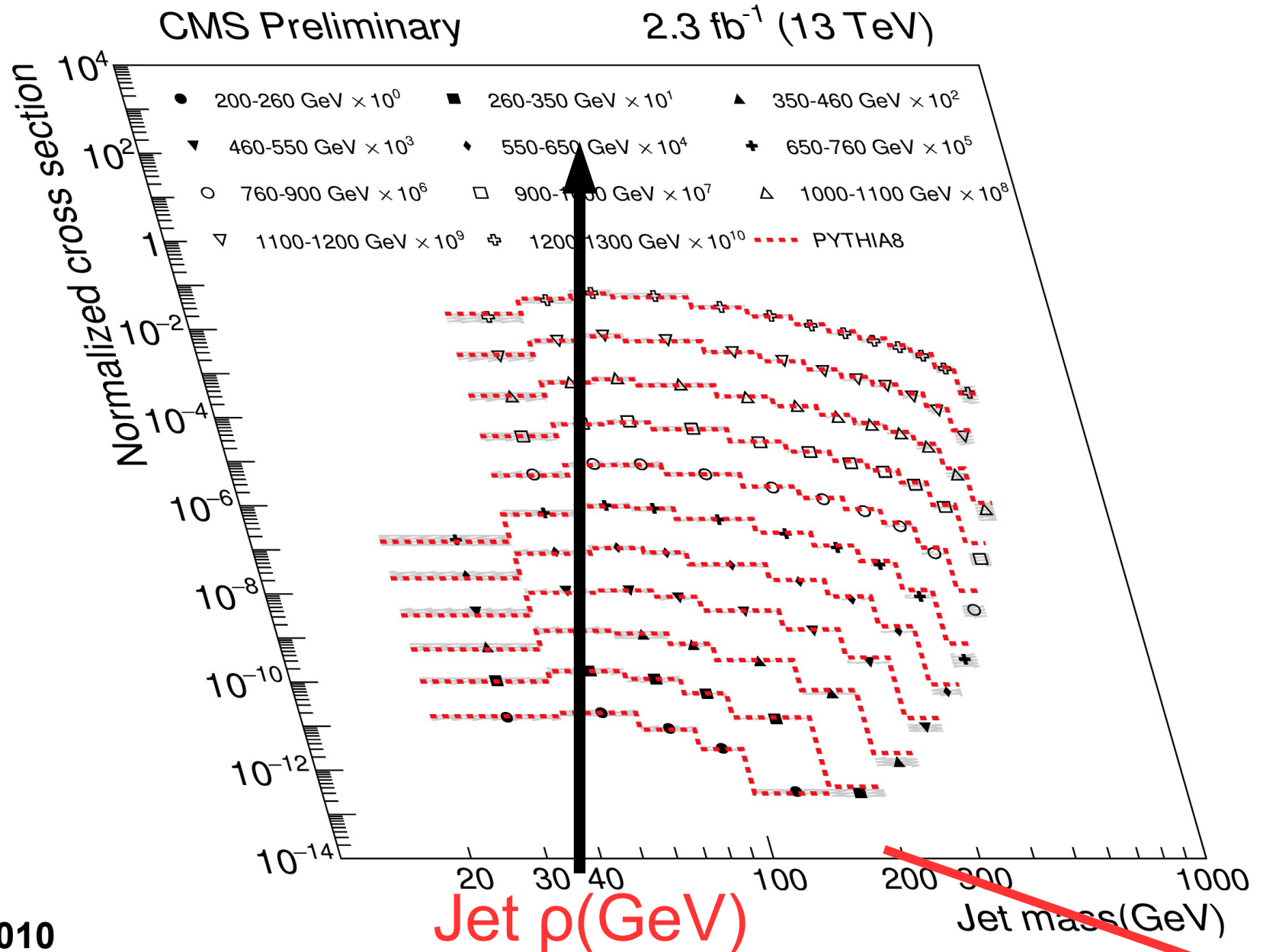
Evolution of Mass Peak

Clear trend in
observed mass



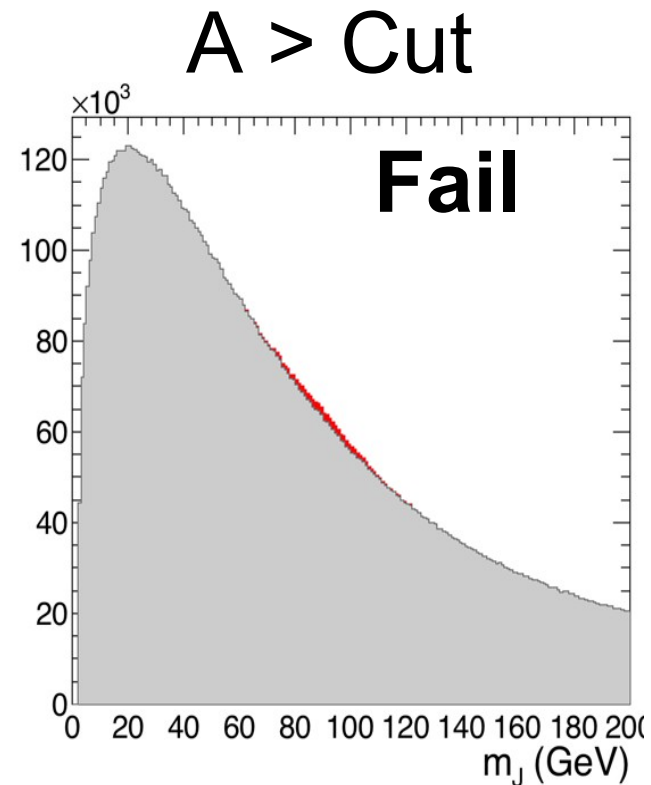
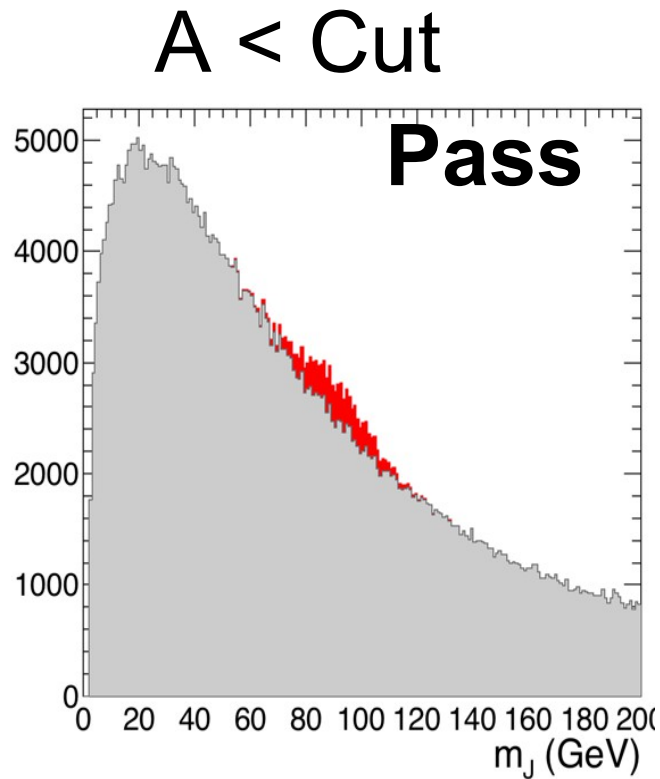
Evolution of Mass Peak

Use QCD to get to modified Jet phase space



One last ingredient

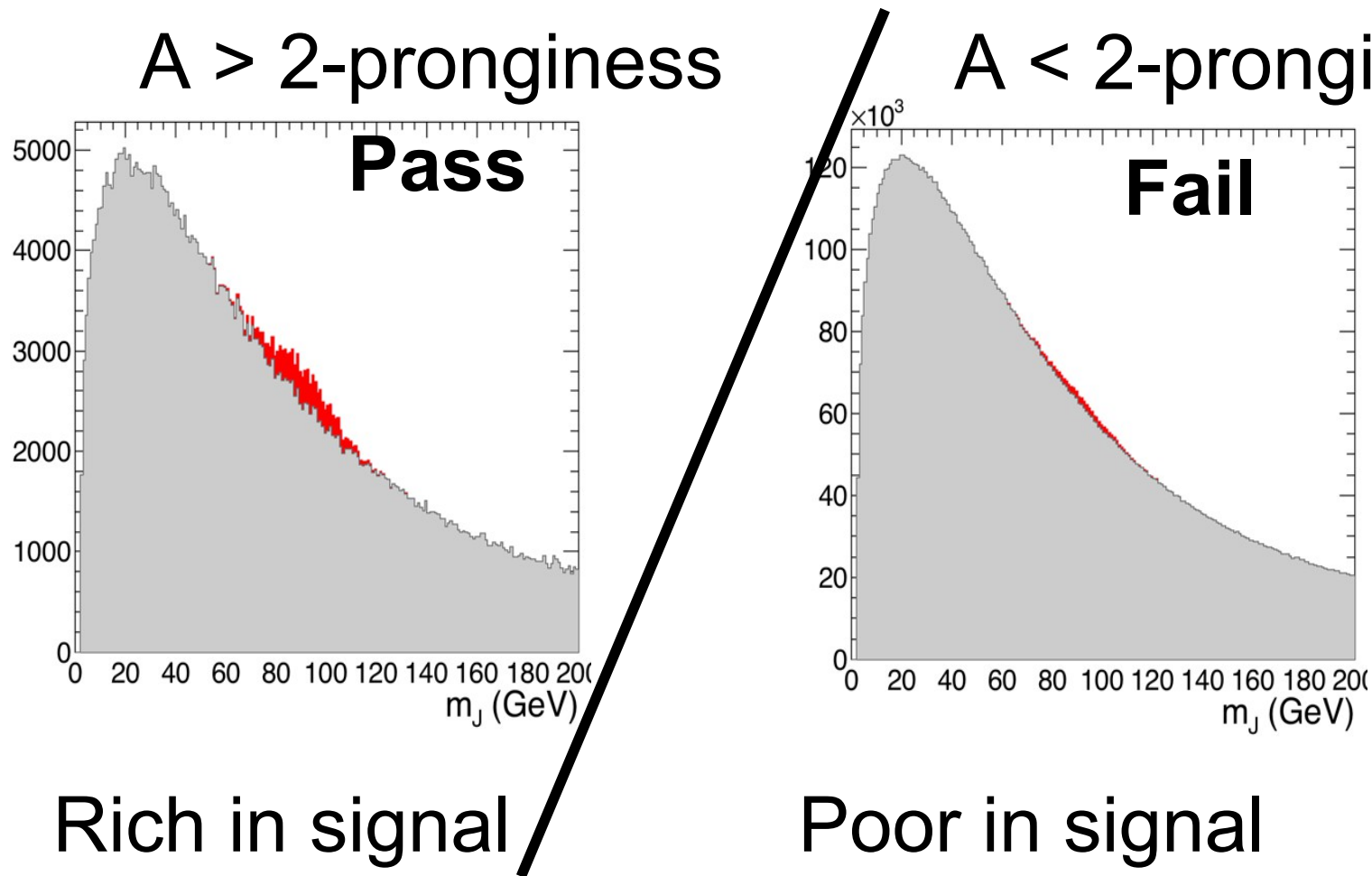
- Cutting on a variable gives us another advantage



We get two regions that we can study

One last ingredient

- Cutting on a variable gives us another advantage

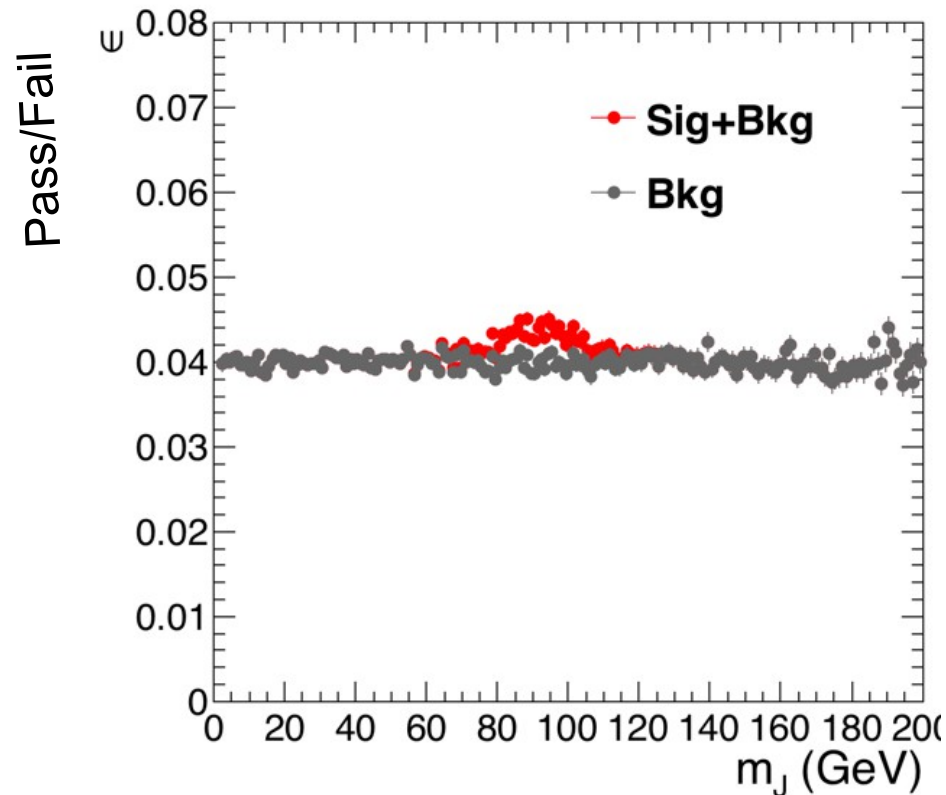


We can merge these regions into one

Pass/Fail Ratio

- Combination of the two yield a single distribution

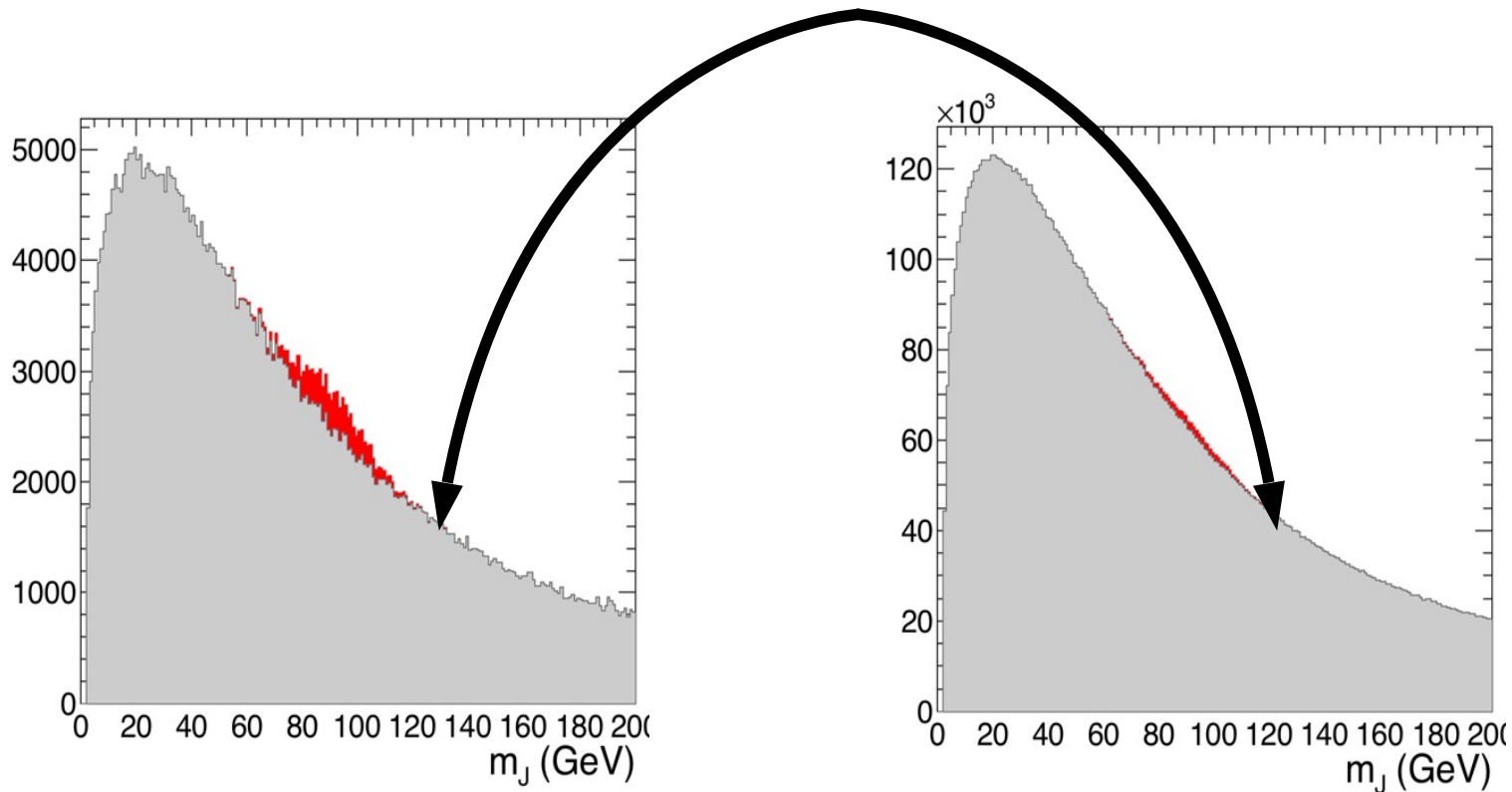
Pass/Fail



This gives us a new combined region

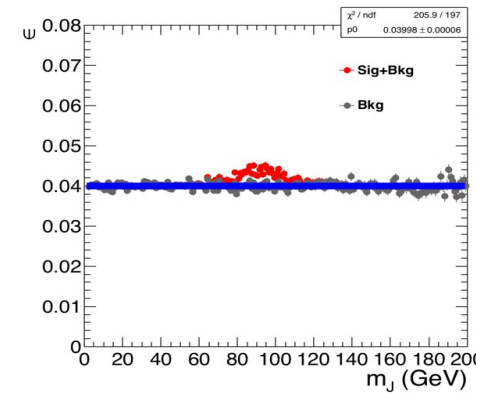
How do you fit the ratio

- Fit both distributions at the same time



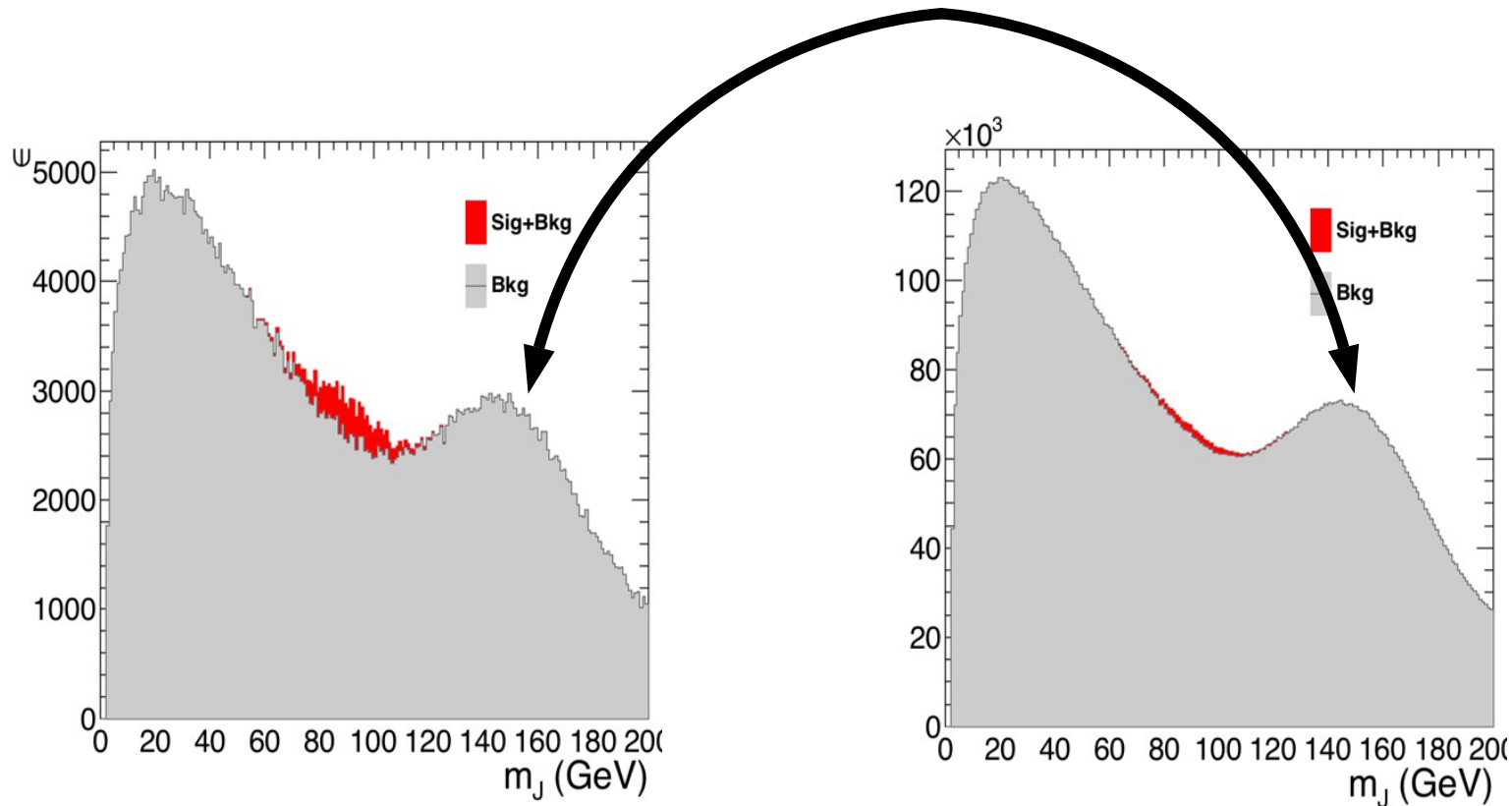
Tie bin in passing to bin in failing

$$\text{Bin}_{\text{Pass}}^i = \varepsilon \text{Bin}_{\text{Fail}}^i$$



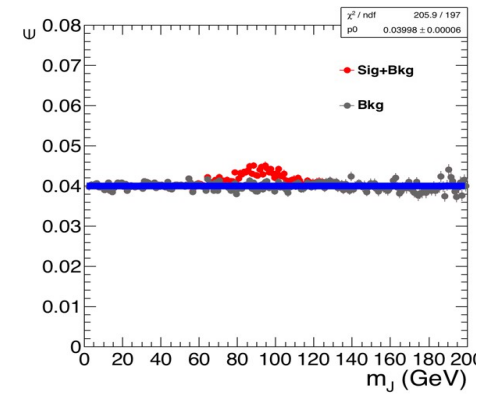
How do you fit the ratio

- Fit both distributions at the same time



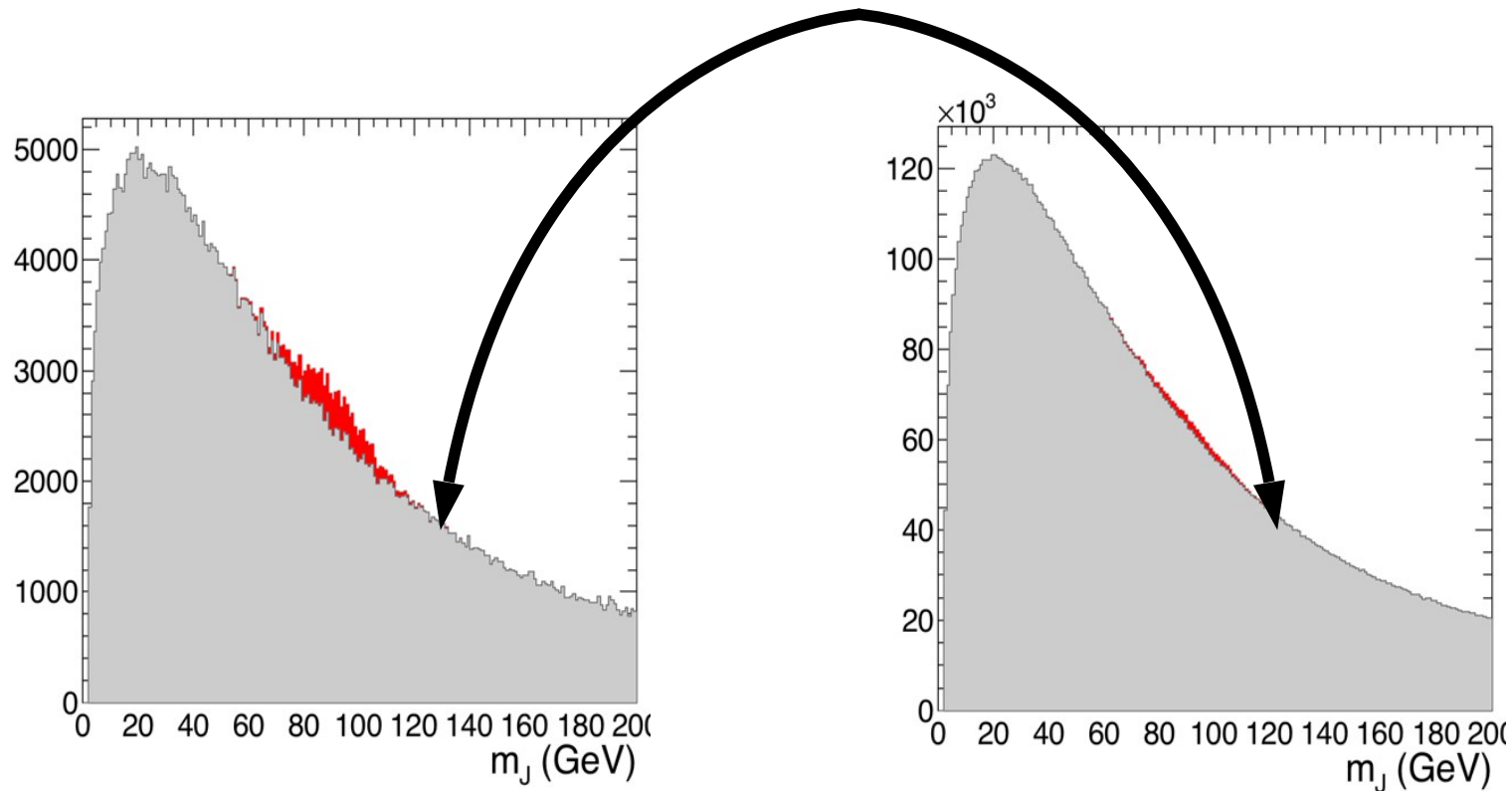
If one bin moves up the other moves up
 Even for weird shapes we fit a flat eff

$$\text{Bin}_{\text{Pass}}^i = \varepsilon \text{Bin}_{\text{Fail}}^i$$

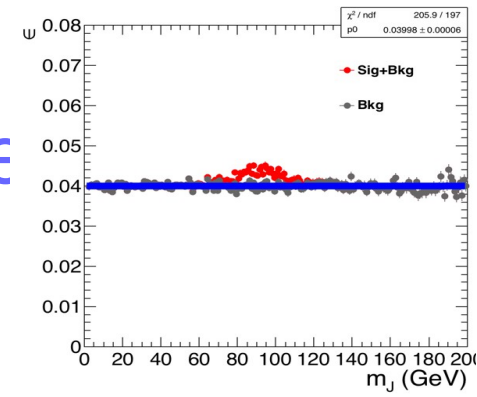


Forcing the ratio flat

- With a large MC sample we can make this flat

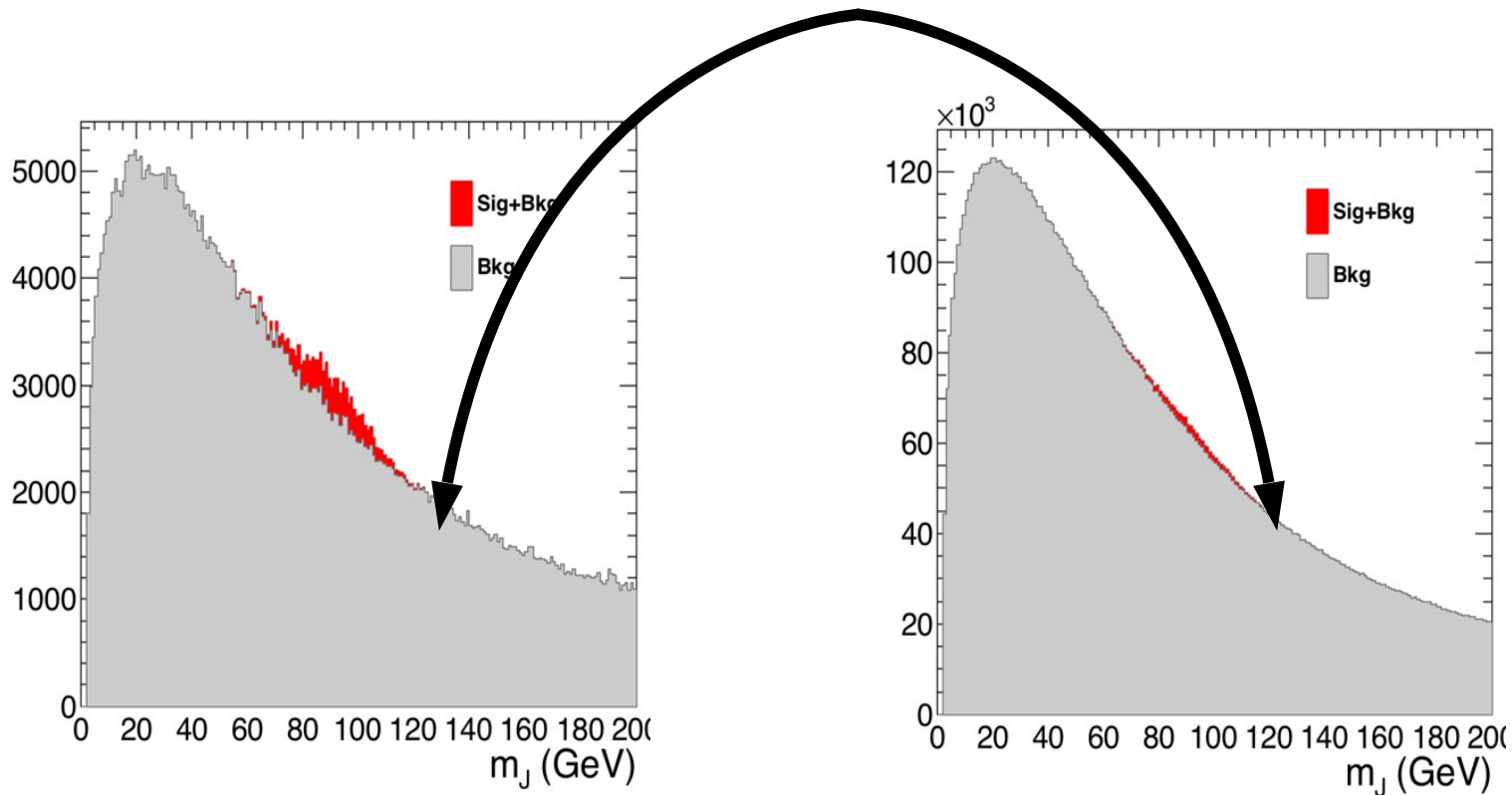


Can construct a substructure observable to have a flat efficiency over mass

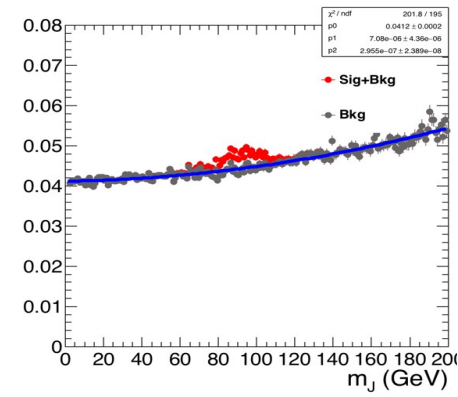


In data

- We can tune the efficiency to be flat in MC

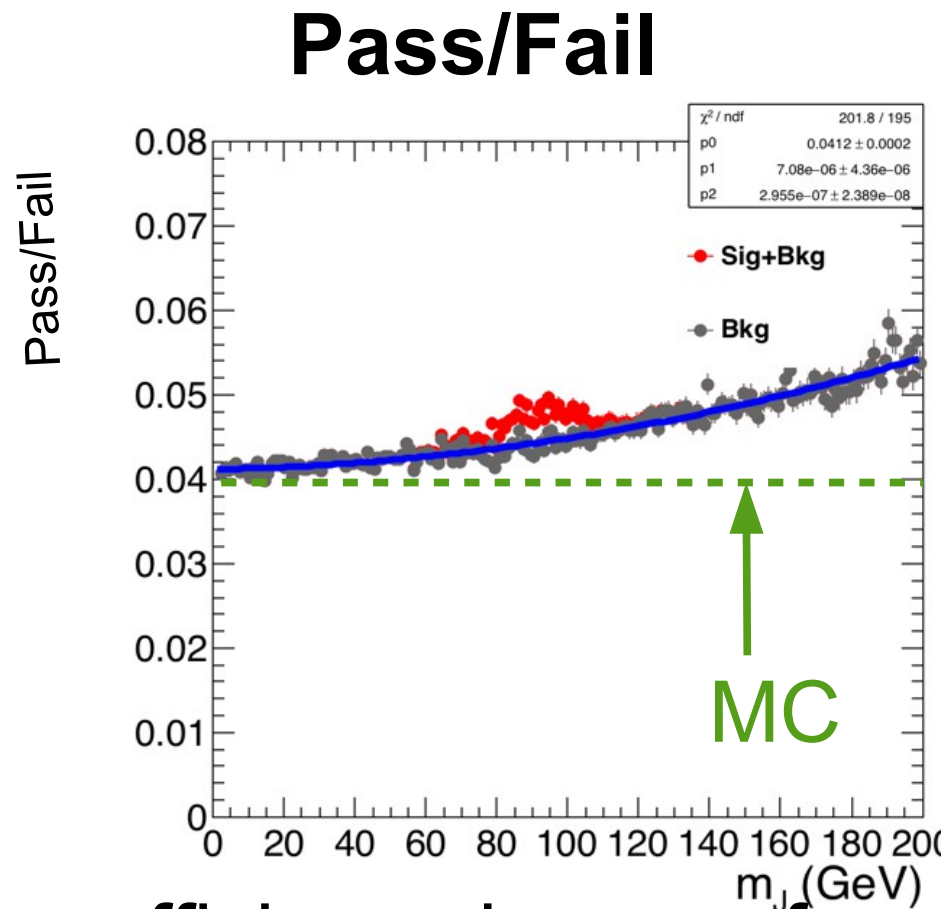


However in data it might not be flat
 We can fit for this deviation over mass



Pass/Fail Ratio

- Parametrize data/MC deviations with function

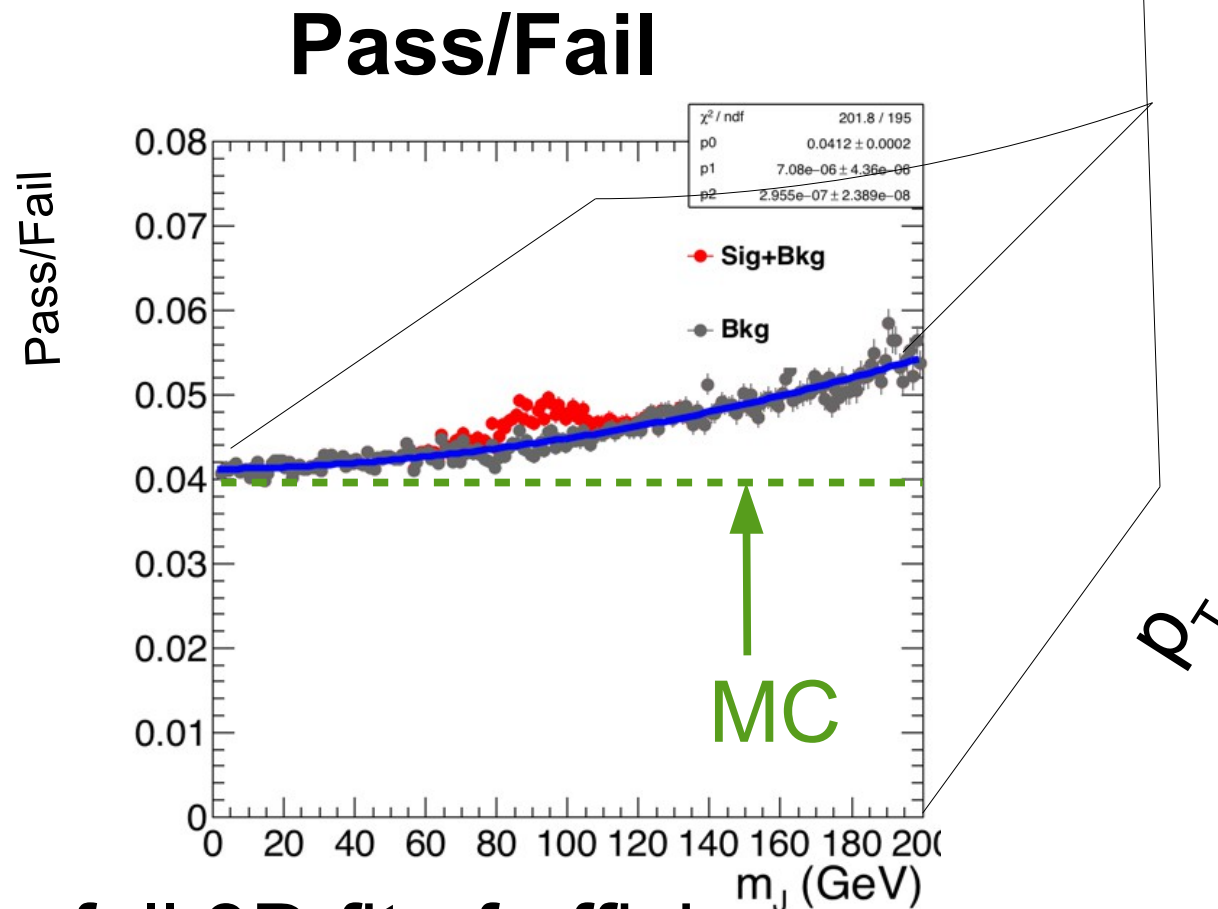


Parametrize efficiency by a surface in ρ and p_T

$$N_{\text{pass}}^{\text{QCD}}(m_{\text{SD}}, p_T) = R_{\text{p/f}}(\rho, p_T) \times N_{\text{fail}}^{\text{QCD}}(m_{\text{SD}}, p_T)$$

Pass/Fail Ratio

- Perform fit in 3 dimensional fit against p_T



Final fit is a full 3D fit of efficiency ρ vs p_T

$$N_{\text{pass}}^{\text{QCD}}(m_{\text{SD}}, p_T) = R_{\text{p/f}}(\rho, p_T) \times N_{\text{fail}}^{\text{QCD}}(m_{\text{SD}}, p_T)$$

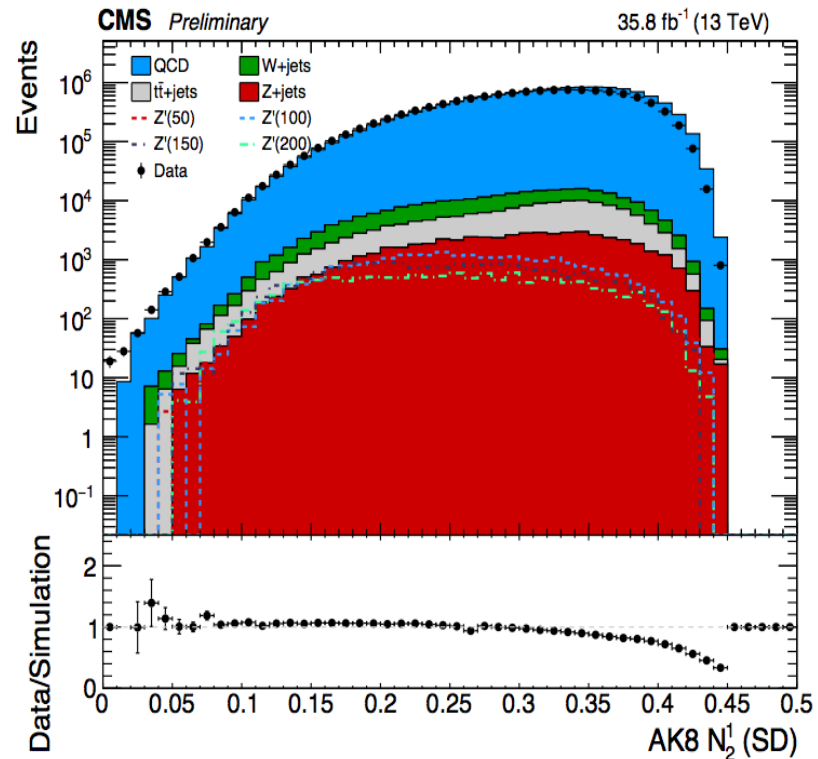
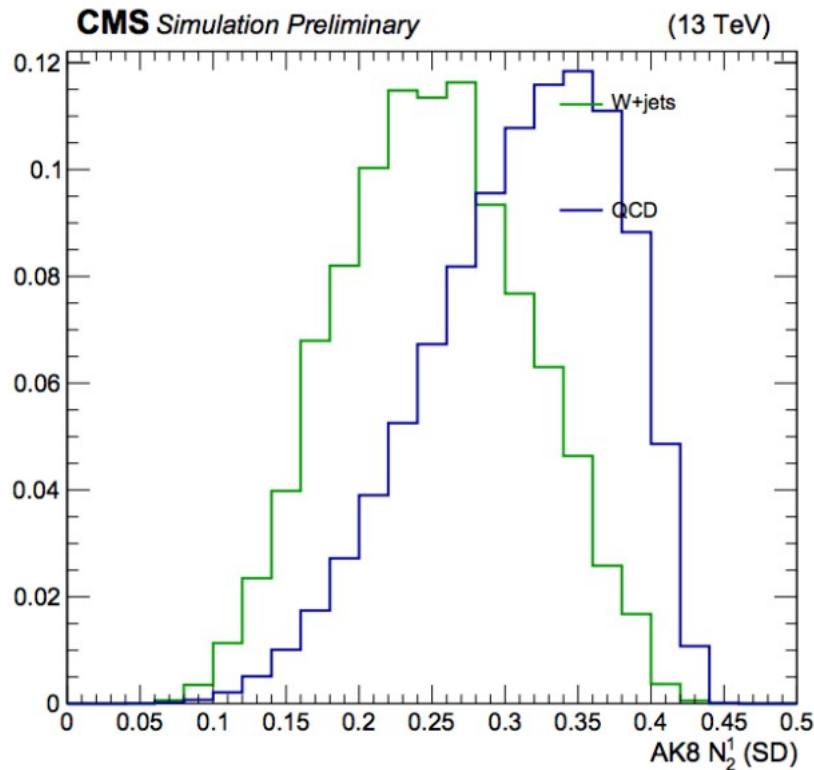
New Substructure Observables

$$N_2(\beta) = \frac{2e_3^\beta}{(e_2^\beta)^2}$$

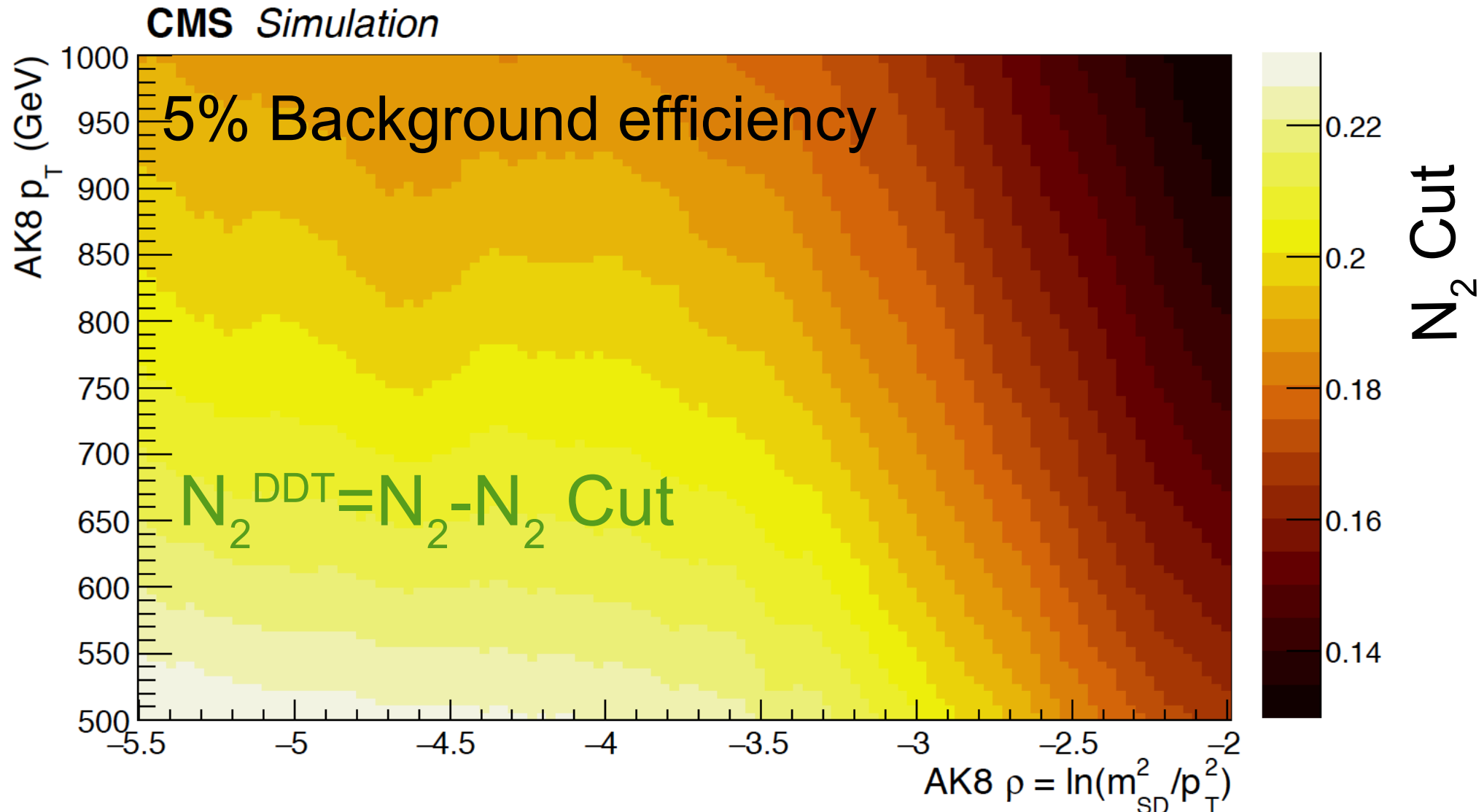
$$1e_2^\beta = e_2^\beta = \sum_{1 \leq i < j \leq n_l} z_i z_j \Delta R_{ij}^\beta \quad z_i \equiv \frac{p_{Ti}}{\sum_{j \in \text{jet}} p_{Tj}}$$

$$2e_3^\beta = \sum_{1 \leq i < j < k \leq n_l} z_i z_j z_k \min\{\Delta R_{ij}^\beta \Delta R_{ik}^\beta, \Delta R_{ij}^\beta \Delta R_{jk}^\beta, \Delta R_{ik}^\beta \Delta R_{jk}^\beta\}$$

Using AK8 PUPPI jets

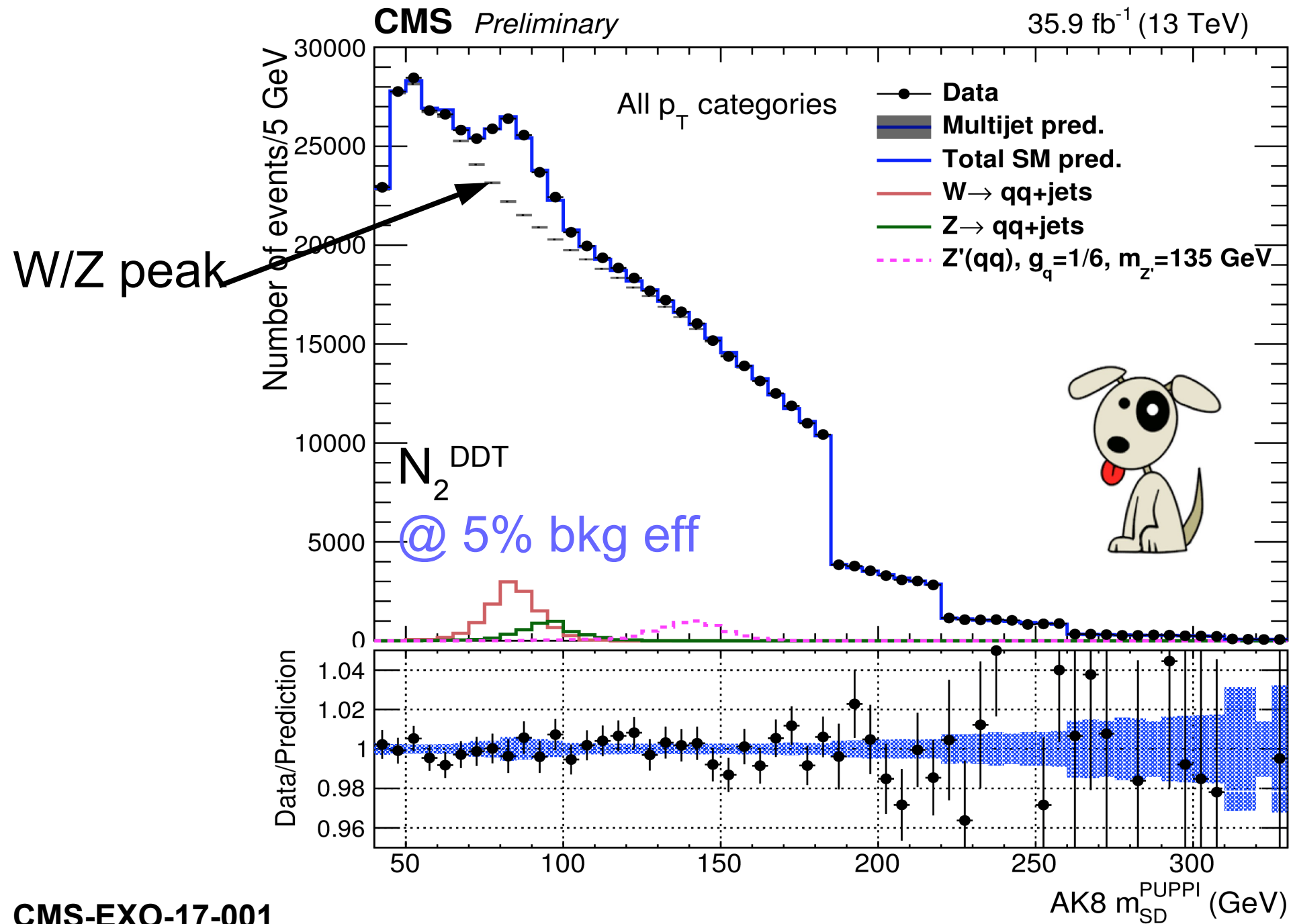


Modify N_2 to have a flat bkg efficiency



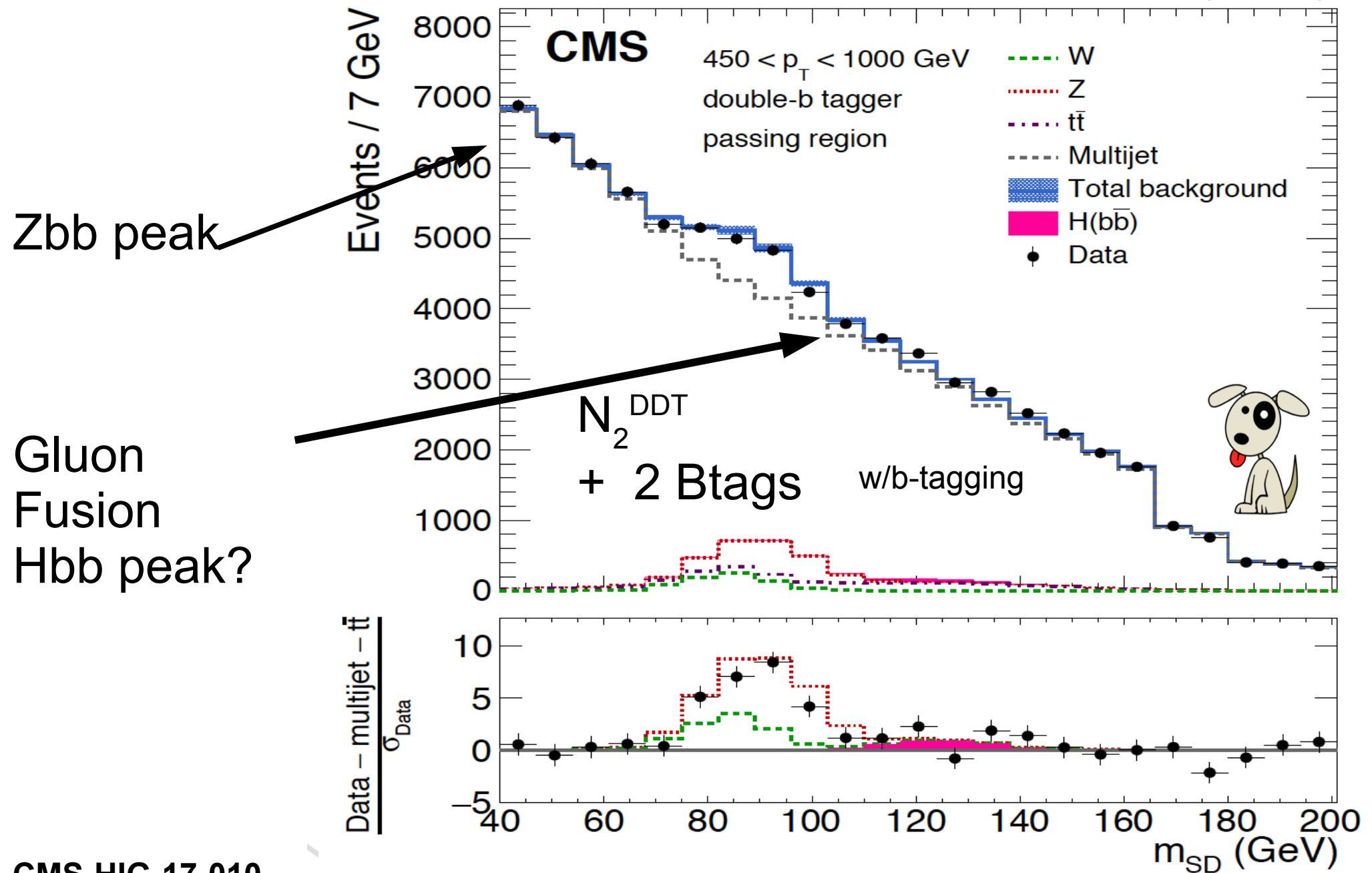
Use the k-Nearest Neighbor approach to determine N_2 cut

Two pronged Jets



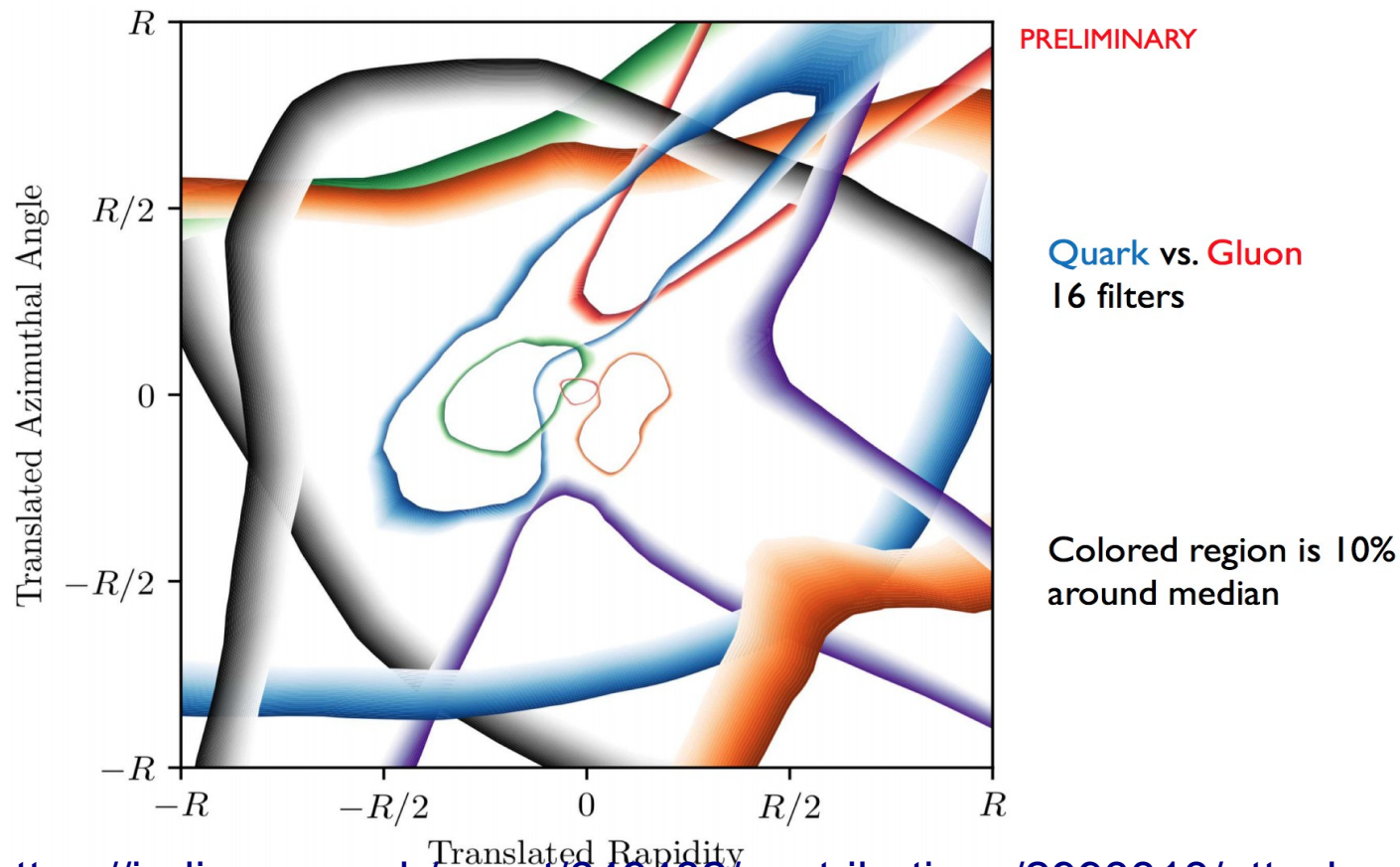
Two pronged Jets+2 b-tags

35.9 fb⁻¹ (13 TeV)



Follow up

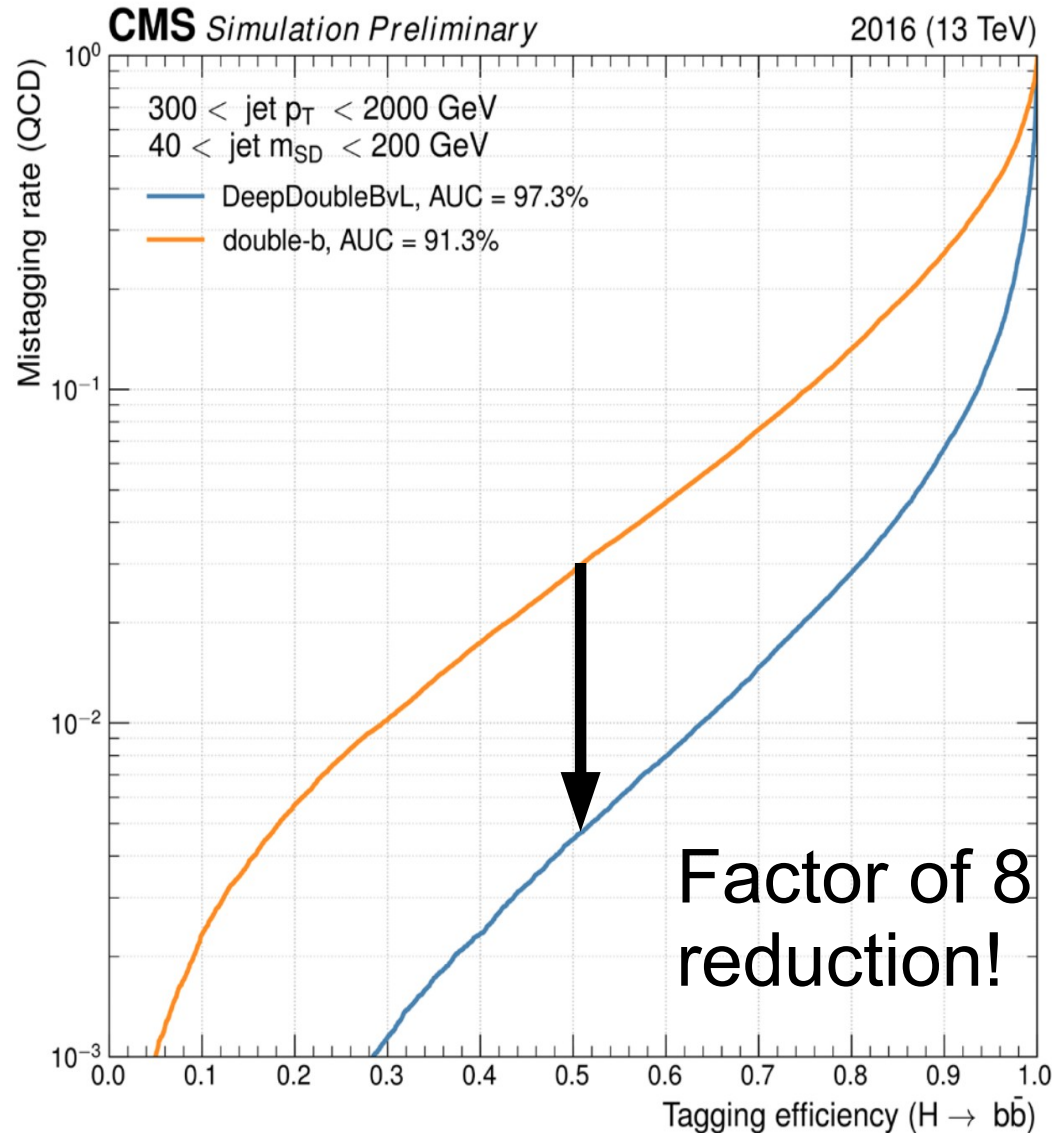
- Where can you improve the analysis?
 - In pp, people are going full deep learning



https://indico.cern.ch/event/649482/contributions/2993313/attachments/1689067/2717141/PTK_B

P. Komiskie, E. Metodiev, J. Thaler Boost 2018 (last week!)

What are the prospects?



Machine learning algo:
All the particles in a jet
(PF candidates in CMS
+2ndary vertices)

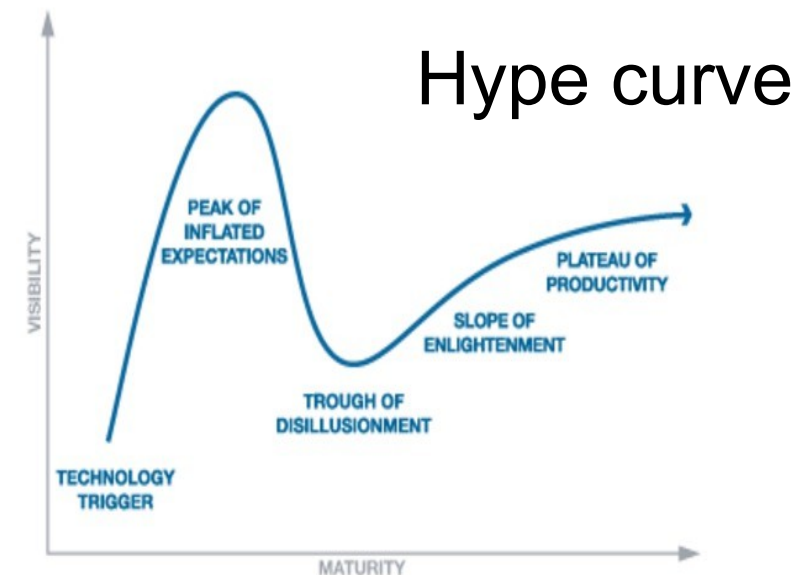
LSTM on all particles
Train for Higgs boson

Impacts of ML can be
very substantial

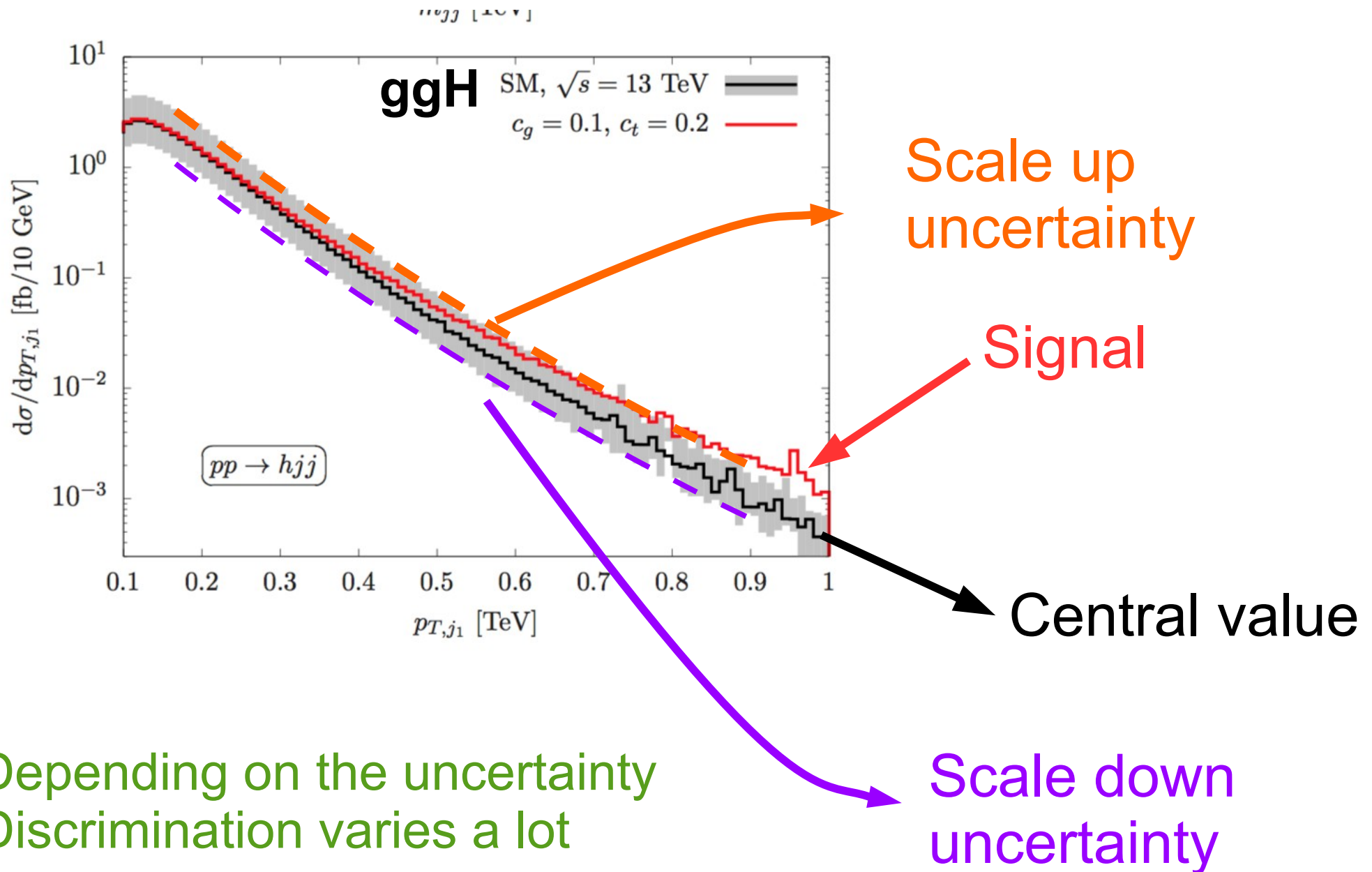
Towards a Realistic use of ML in jets

- Despite the big excitement about ML
 - People are **hesitant** to use these
 - Simply because its **not clear where we are going**
 - Universal thing: is that all ML algos give ~same gain

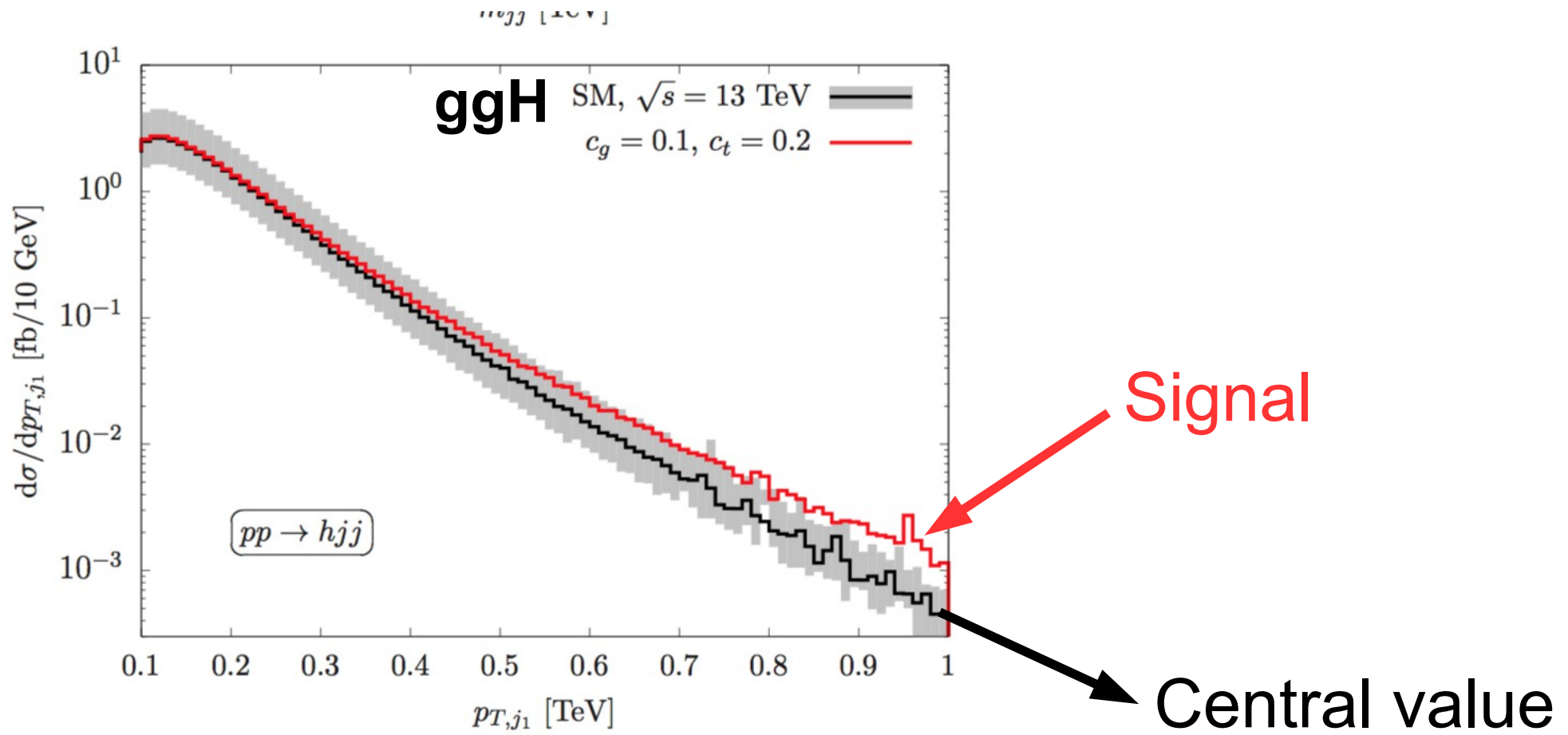
- In lieu of this....
 - Focus has been on physics
 - **How can we restrict MLs**
 - Avoid unphyscial gains
 - **Ensure physics is preserved**
 - Use our bkg methods



Importance of Adversaries

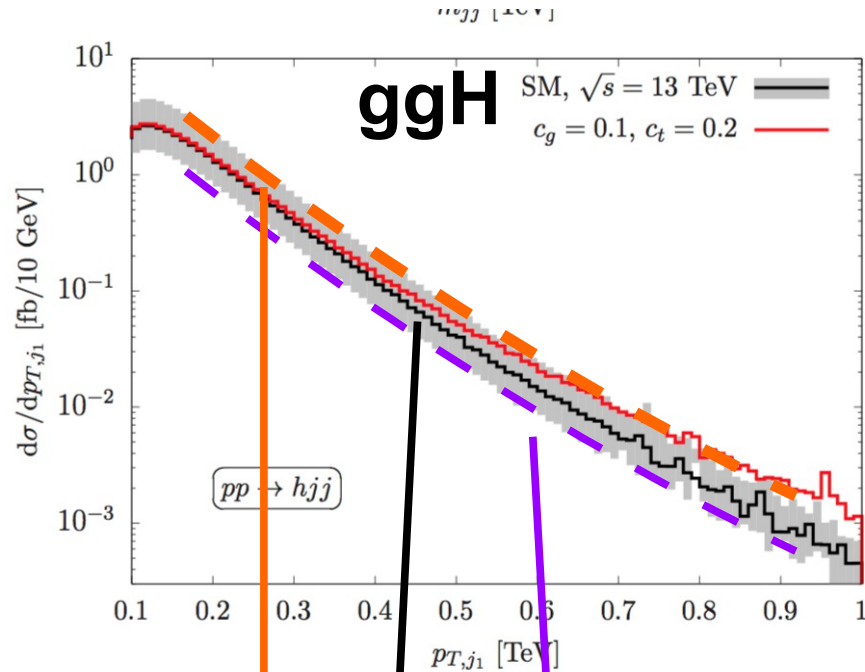


Importance of Adversaries



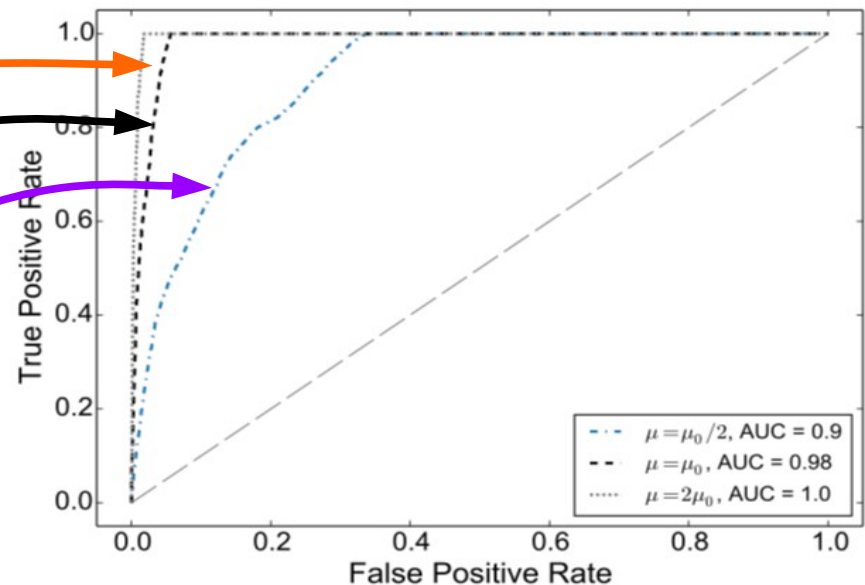
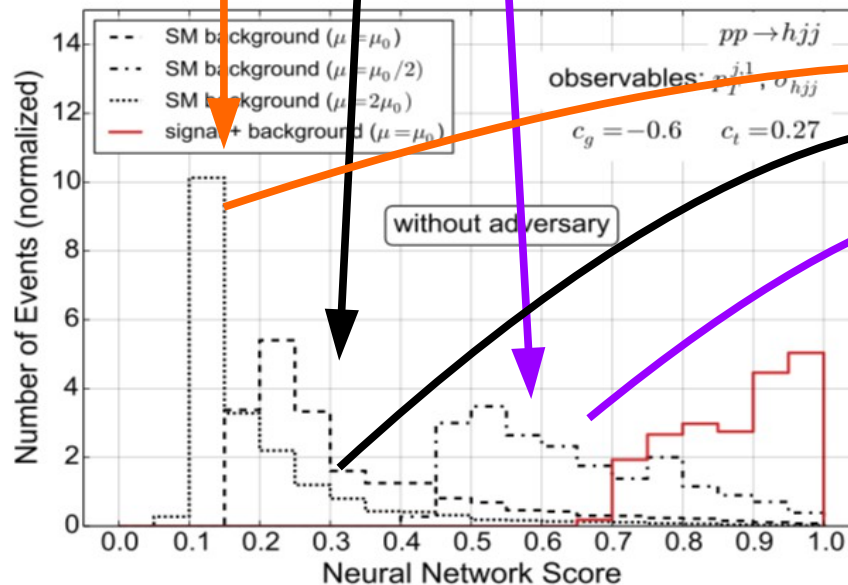
Depending on the uncertainty
 Discrimination varies a lot

Impact on discrimination



Depending on your uncertainty

Can claim very large discrimination performance



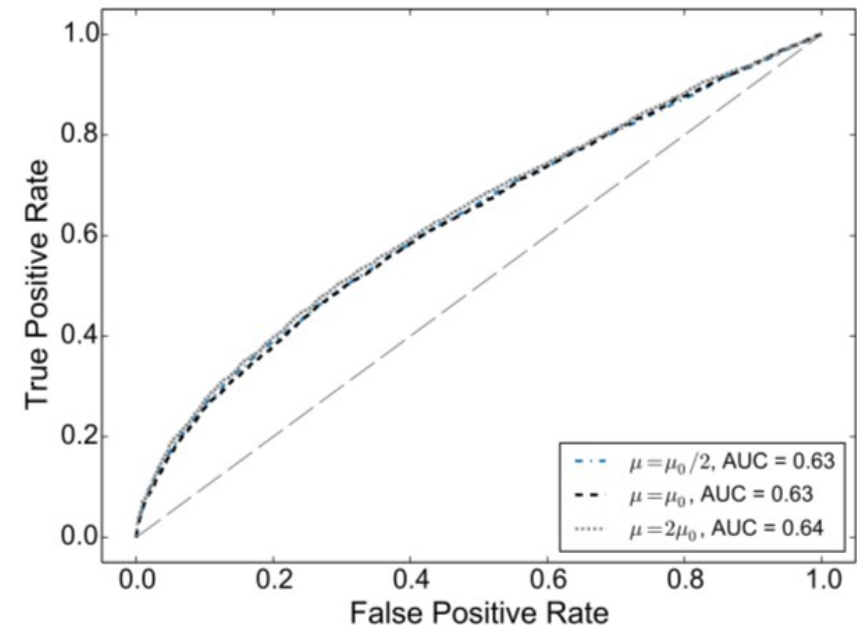
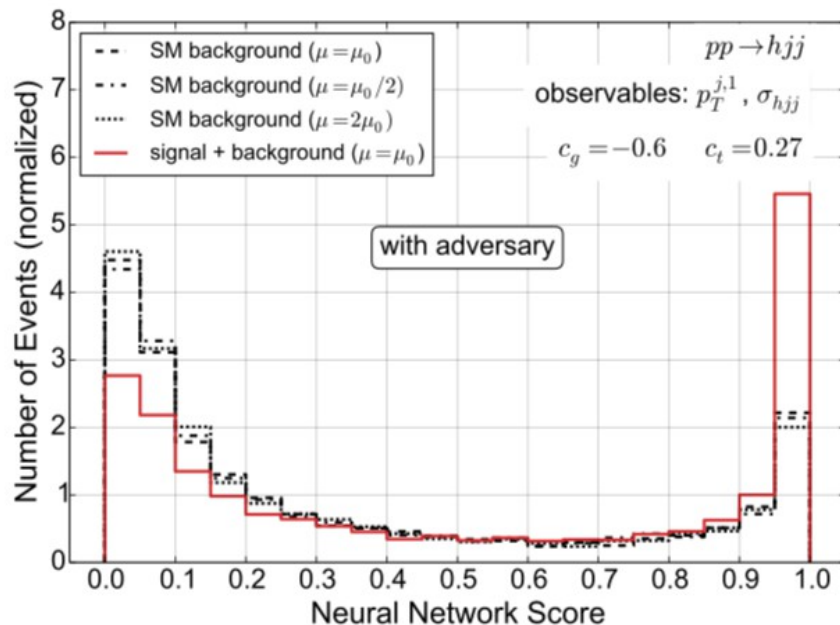
Impact on discrimination

Through the use of adversarial neural networks

Can implicitly build in uncertainties

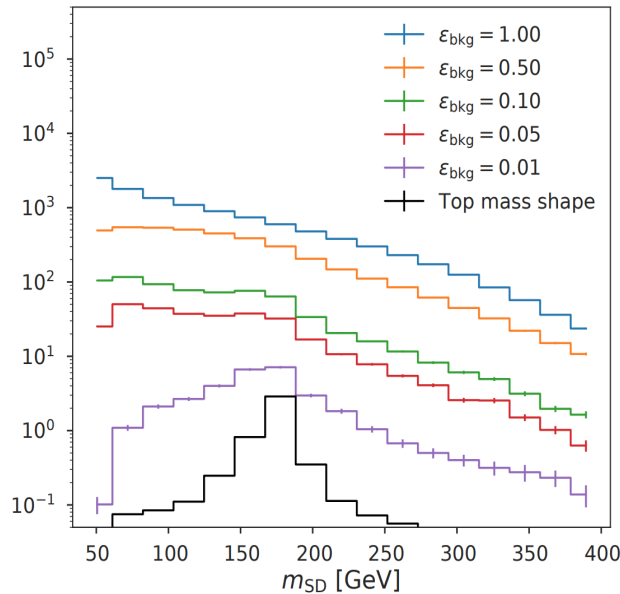
This allows for directed measurements

Preserve ability to account high dimensional inputs

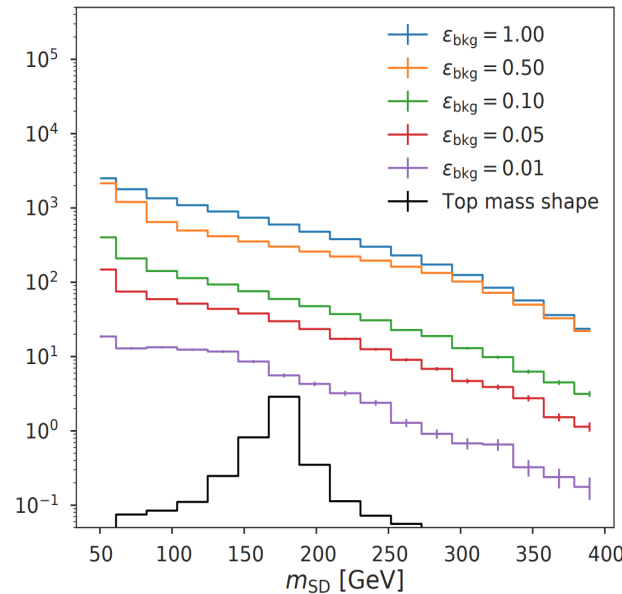


In the context of Jets decorrelation

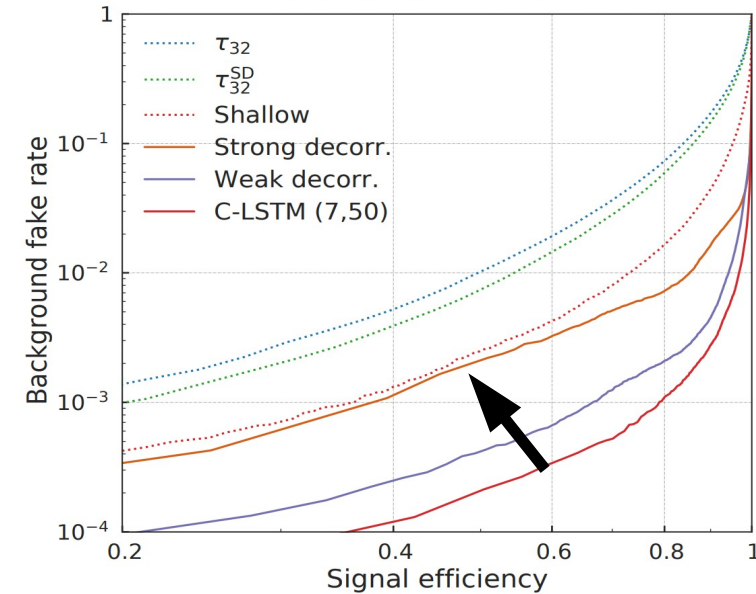
Deep NN



Decorr. deep NN

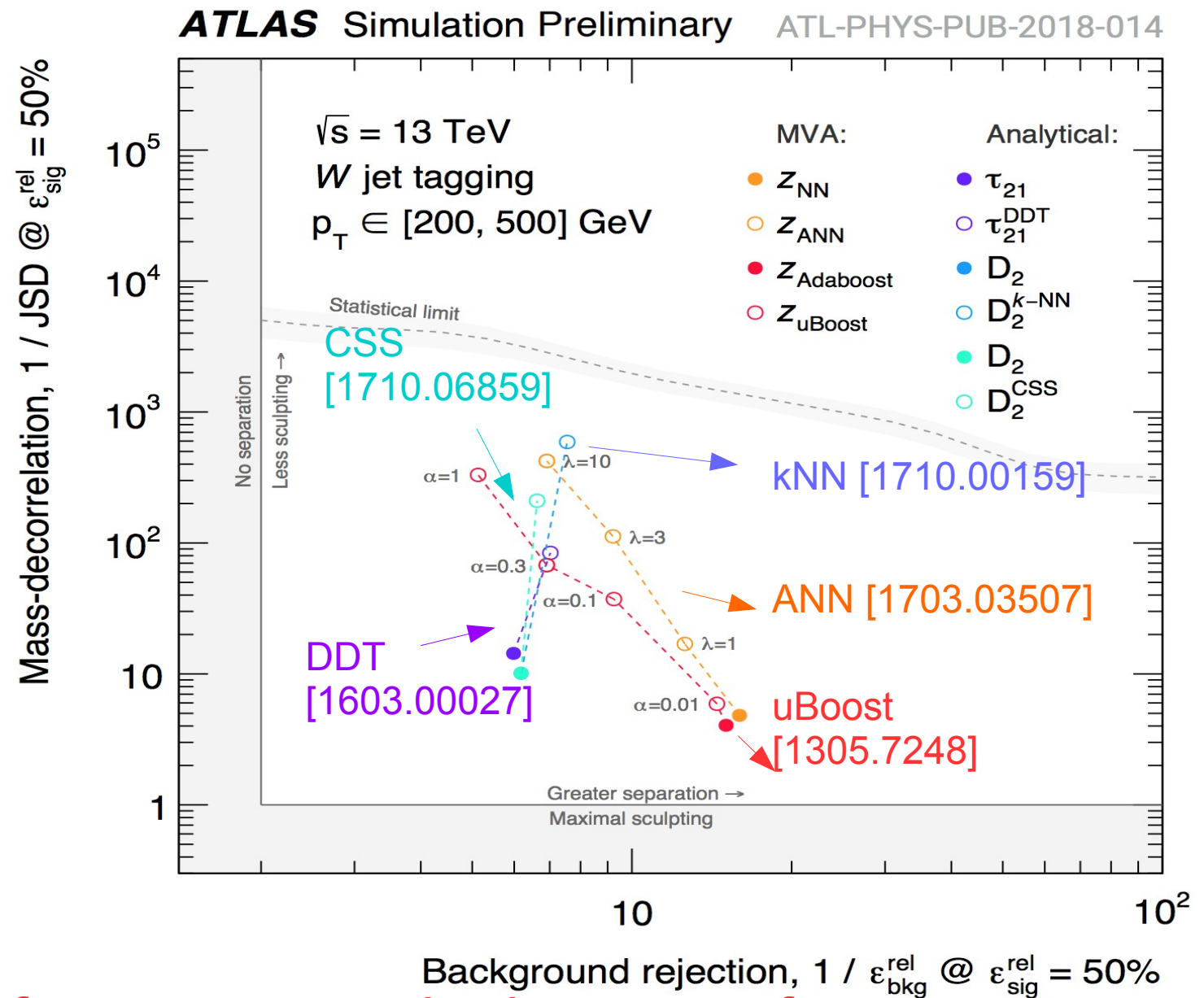


Performance



Following the paradigm of V+jets search
 We build a giant NN (> 100 particles inputs)
 Decorrelate NN against mass
 Scale of decorrelation drives perf.

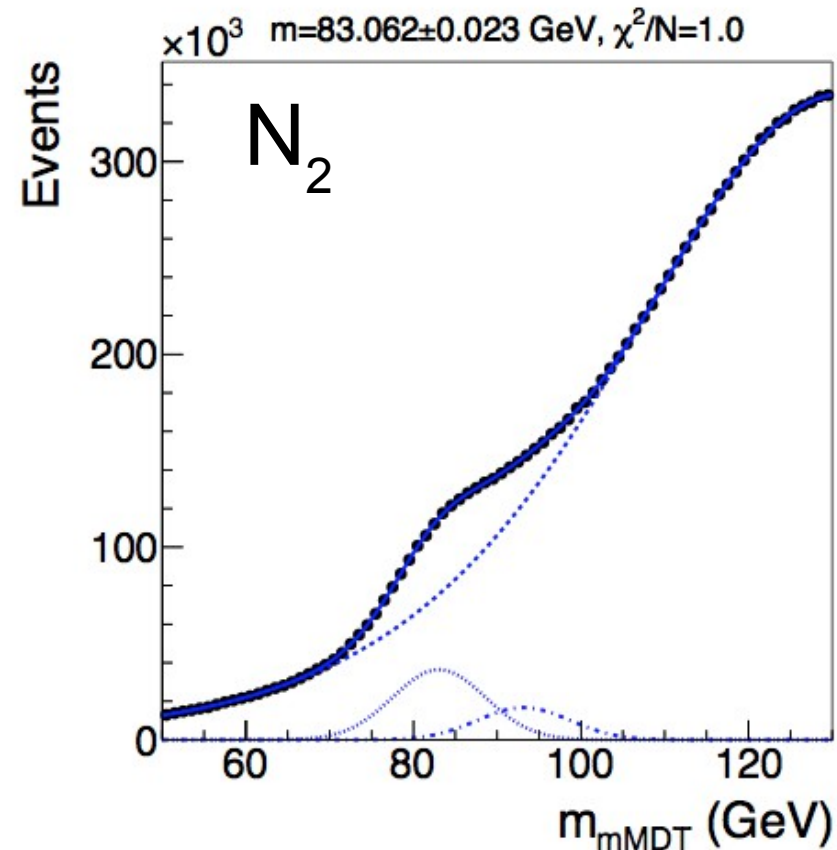
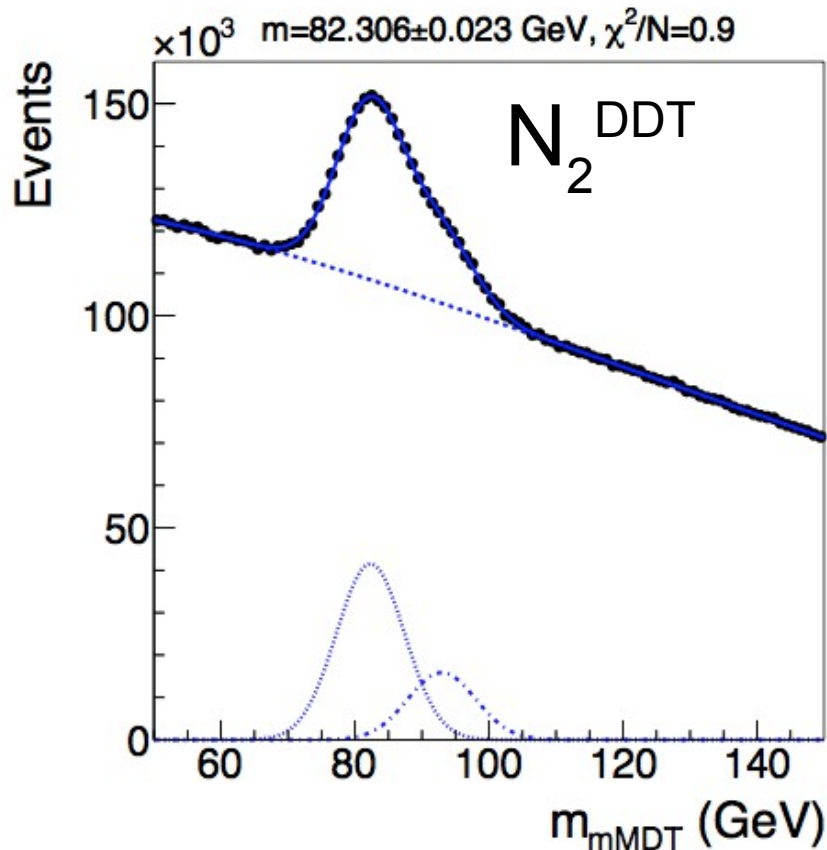
Scanning the decorrelation schemes



Decorrelation of mass seen to be important for new taggers

Can we do a fundamental meas.?

- Would it be possible to apply this to W mass?



With 3/ab and $p_T > 500$ GeV

have a statistical precision of 23 MeV and **systematics?**

Arxiv/1807.07454 (last week)

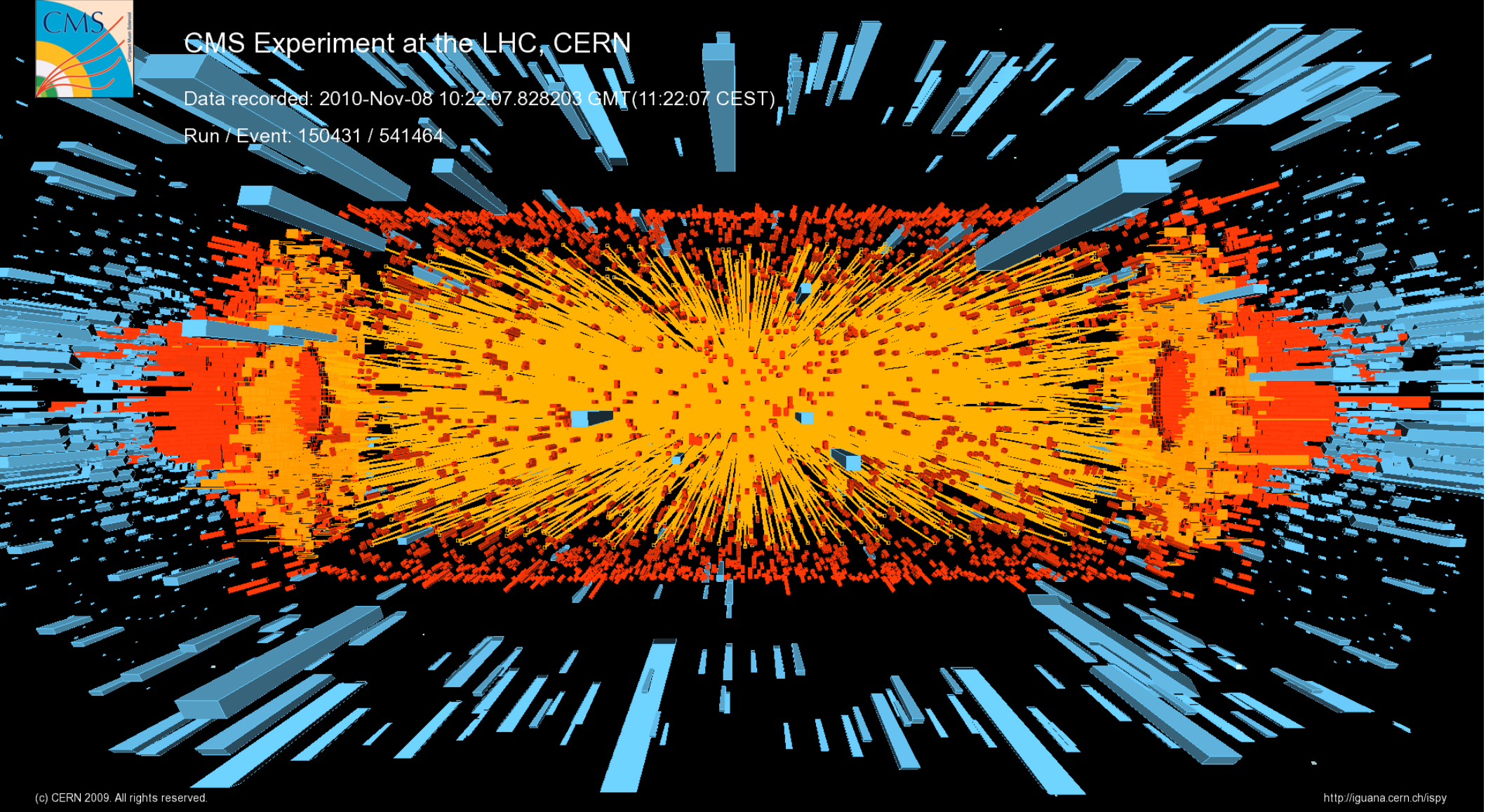
Heavy Ion Collisions



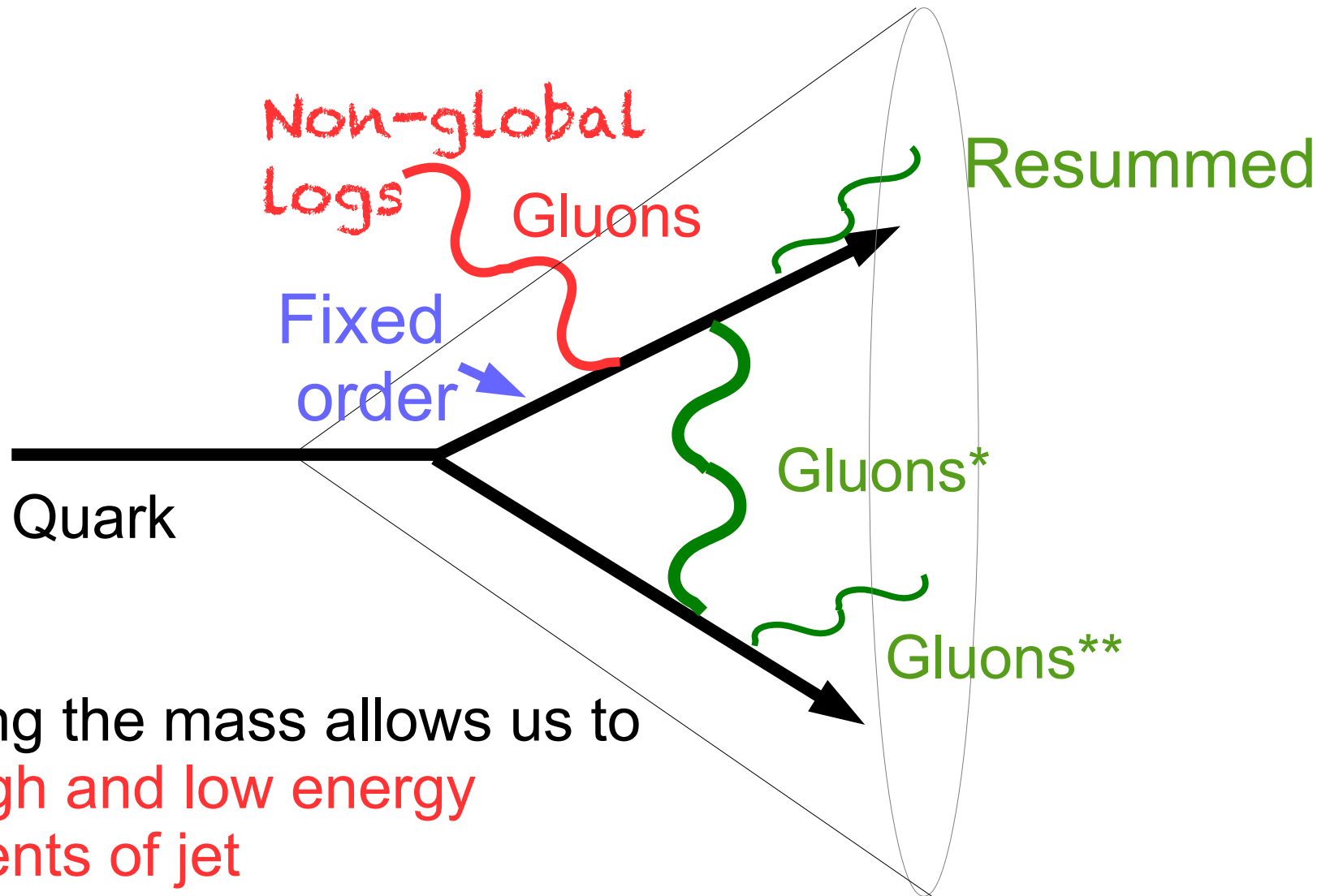
CMS Experiment at the LHC, CERN

Data recorded: 2010-Nov-08 10:22:07.828203 GMT (11:22:07 CEST)

Run / Event: 150431 / 541464

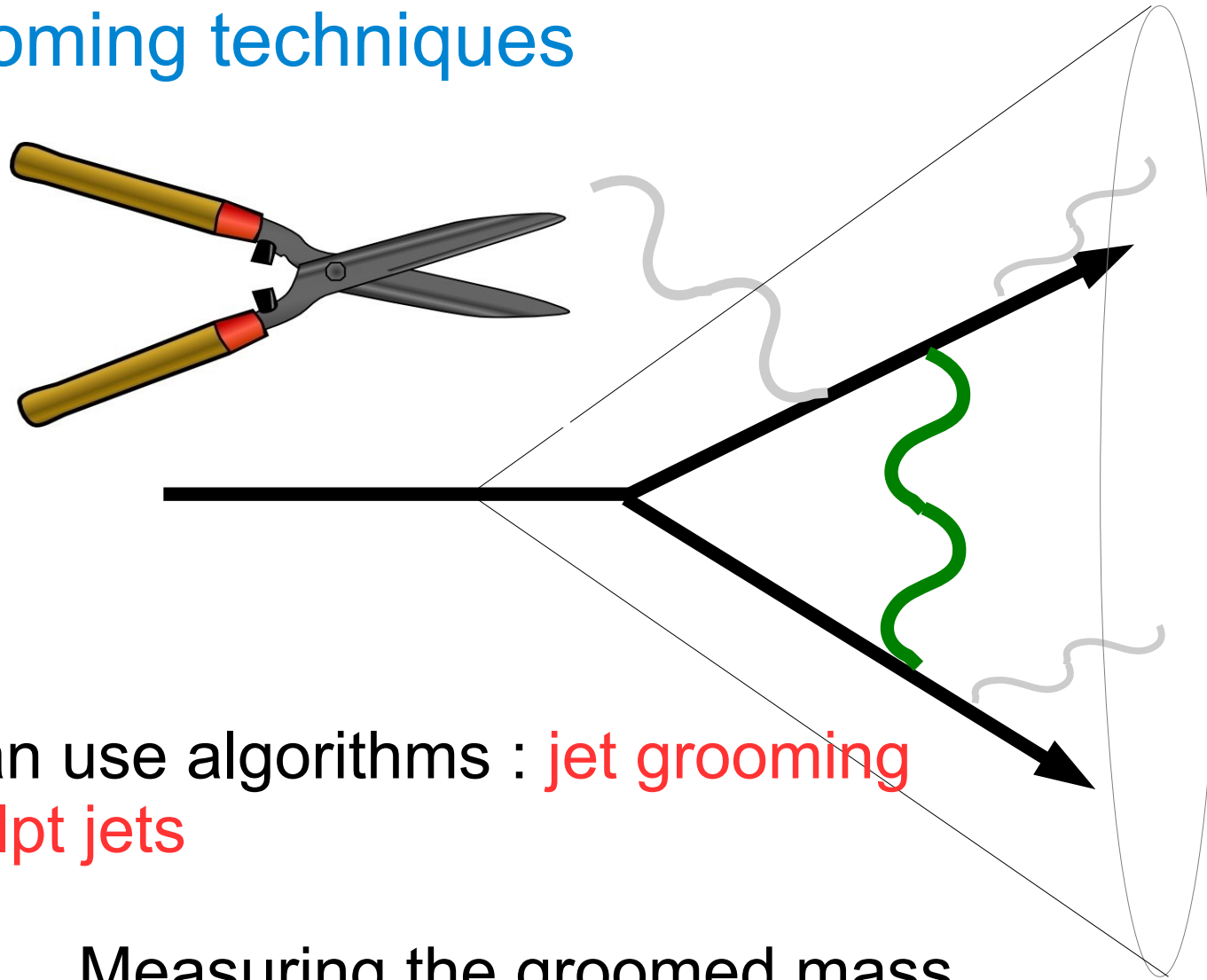


Visualizing Jet Mass



Grooming Jet Mass

Grooming techniques



We can use algorithms : **jet grooming**
to sculpt jets

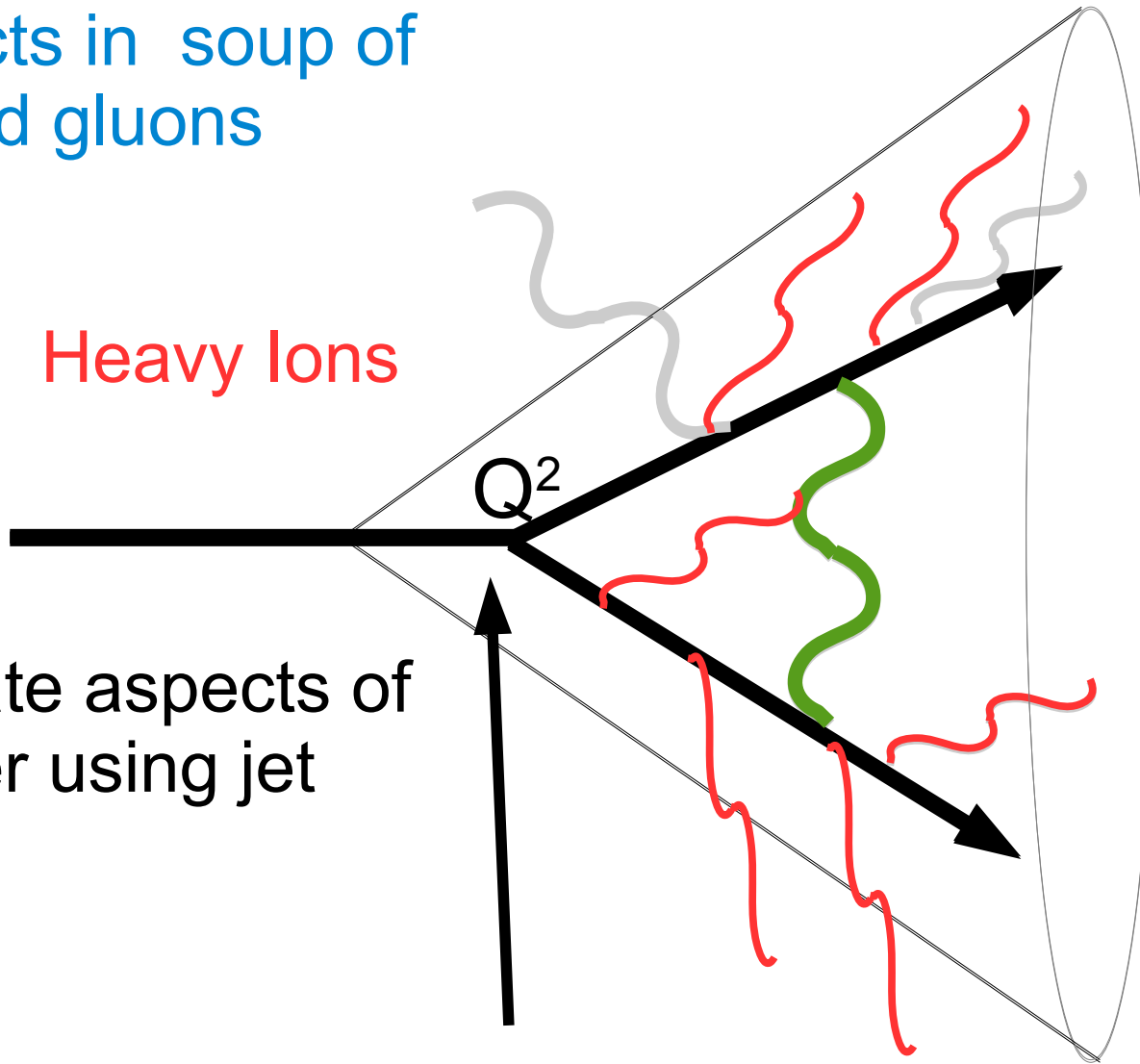
Measuring the groomed mass
allows us to **probe the hard**
component

Heavy Ion Collisions

Jet interacts in soup of quarks and gluons

Heavy Ions

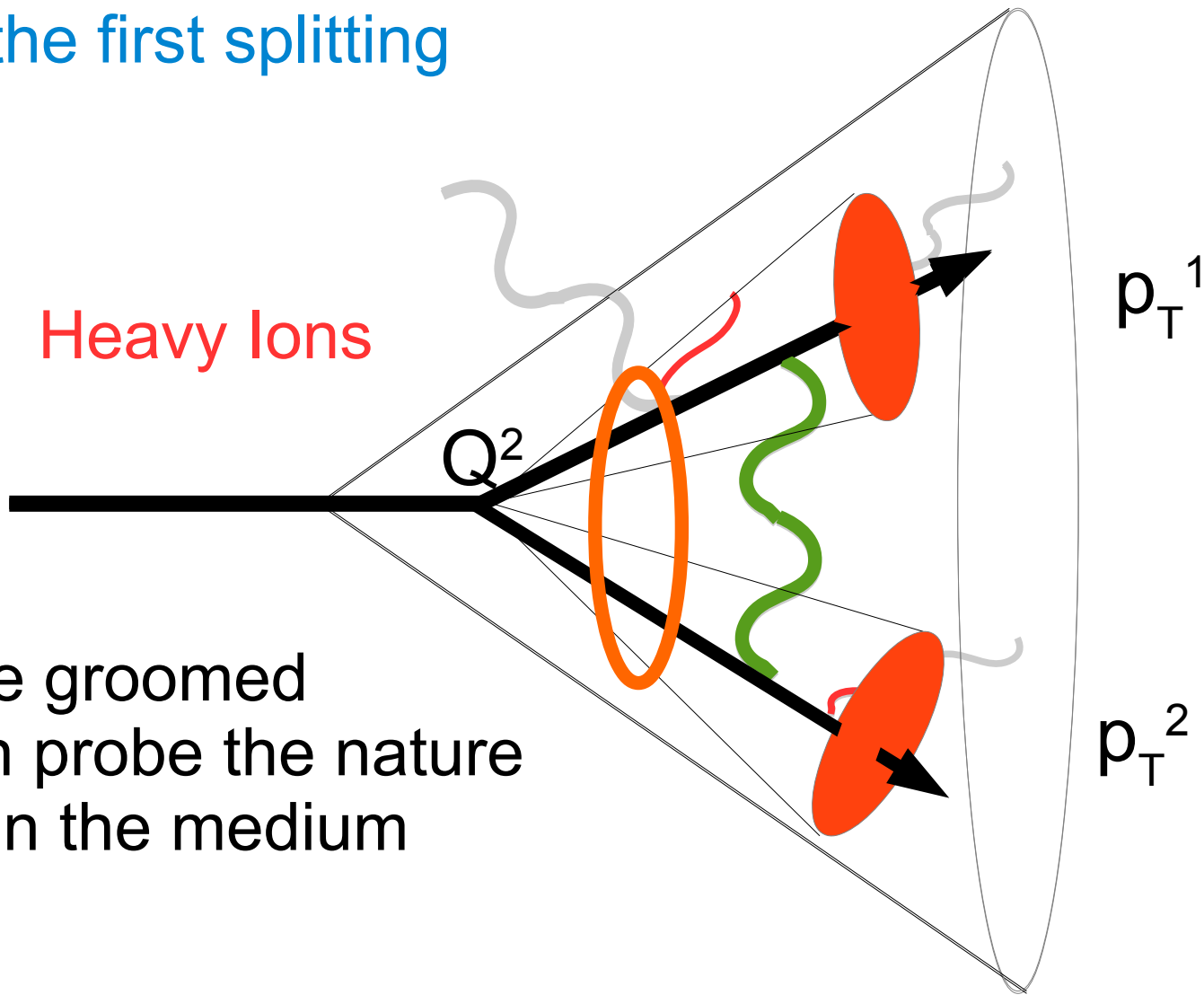
try to isolate aspects of the shower using jet grooming



Properties of the jet allow us to define
Medium modifications

Heavy Ion Collisions

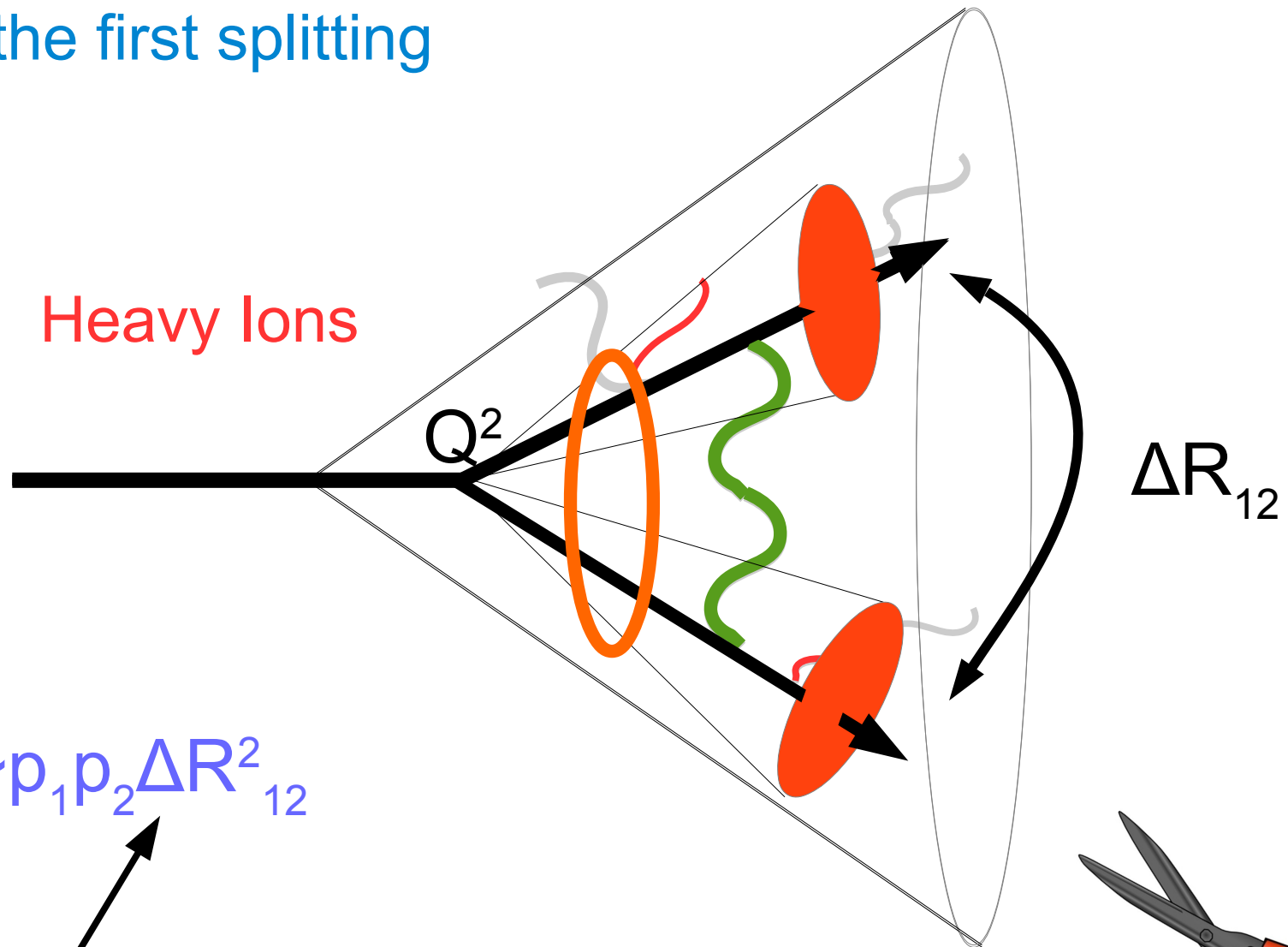
Consider the first splitting
in jet



Isolating the groomed
Subjects can probe the nature
of the loss in the medium

Heavy Ion Collisions

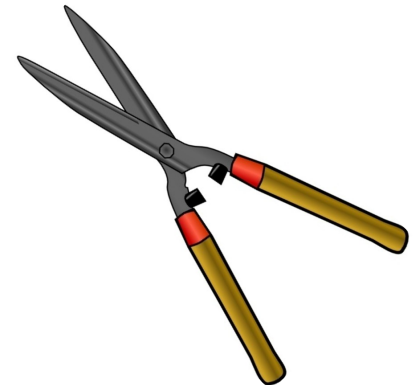
Consider the first splitting
in jet



Jet Mass

$$M^2 \sim p_1 p_2 \Delta R_{12}^2$$

Tells us about the medium

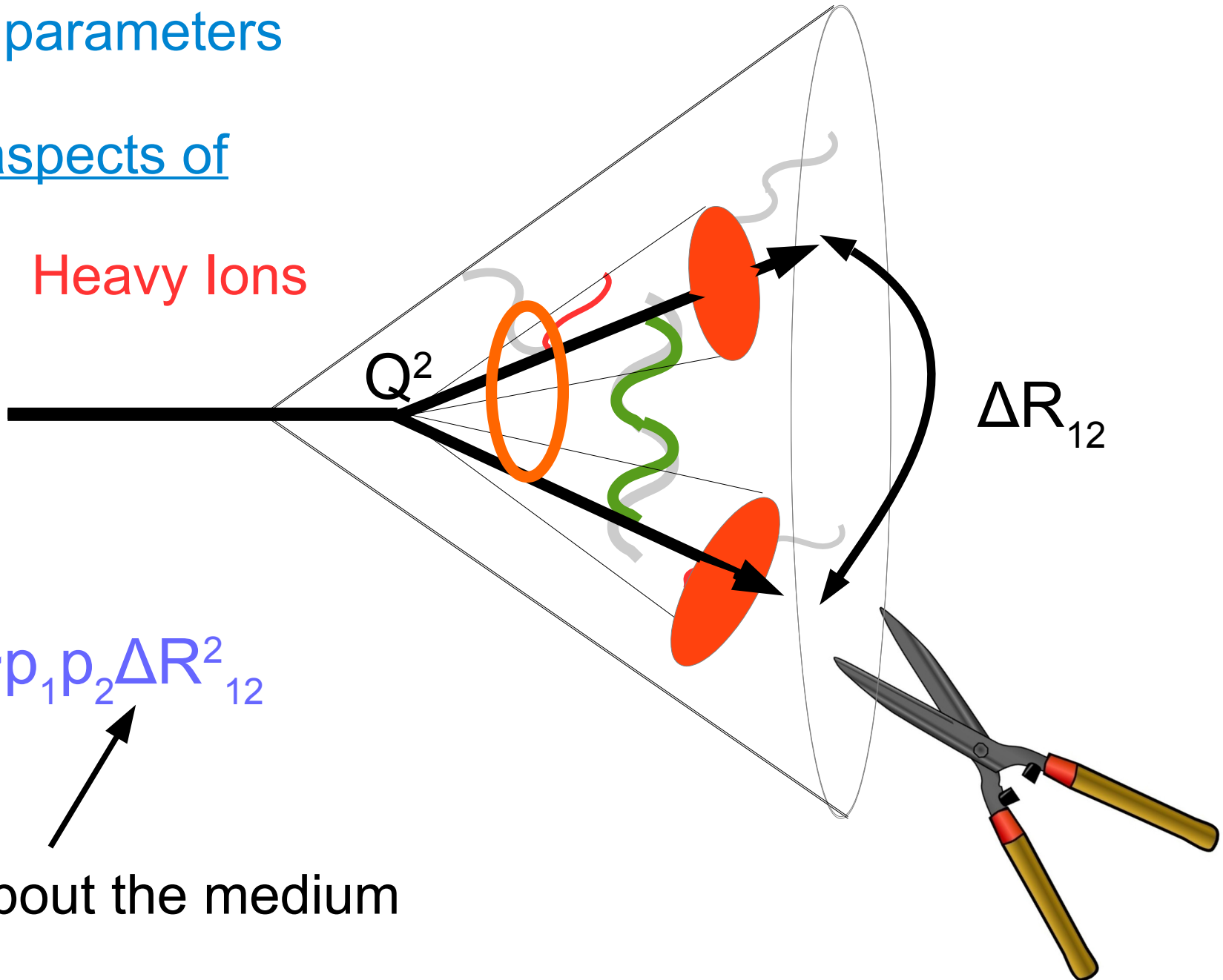


Heavy Ion Collisions

Modifying parameters

Different aspects of medium

Heavy Ions

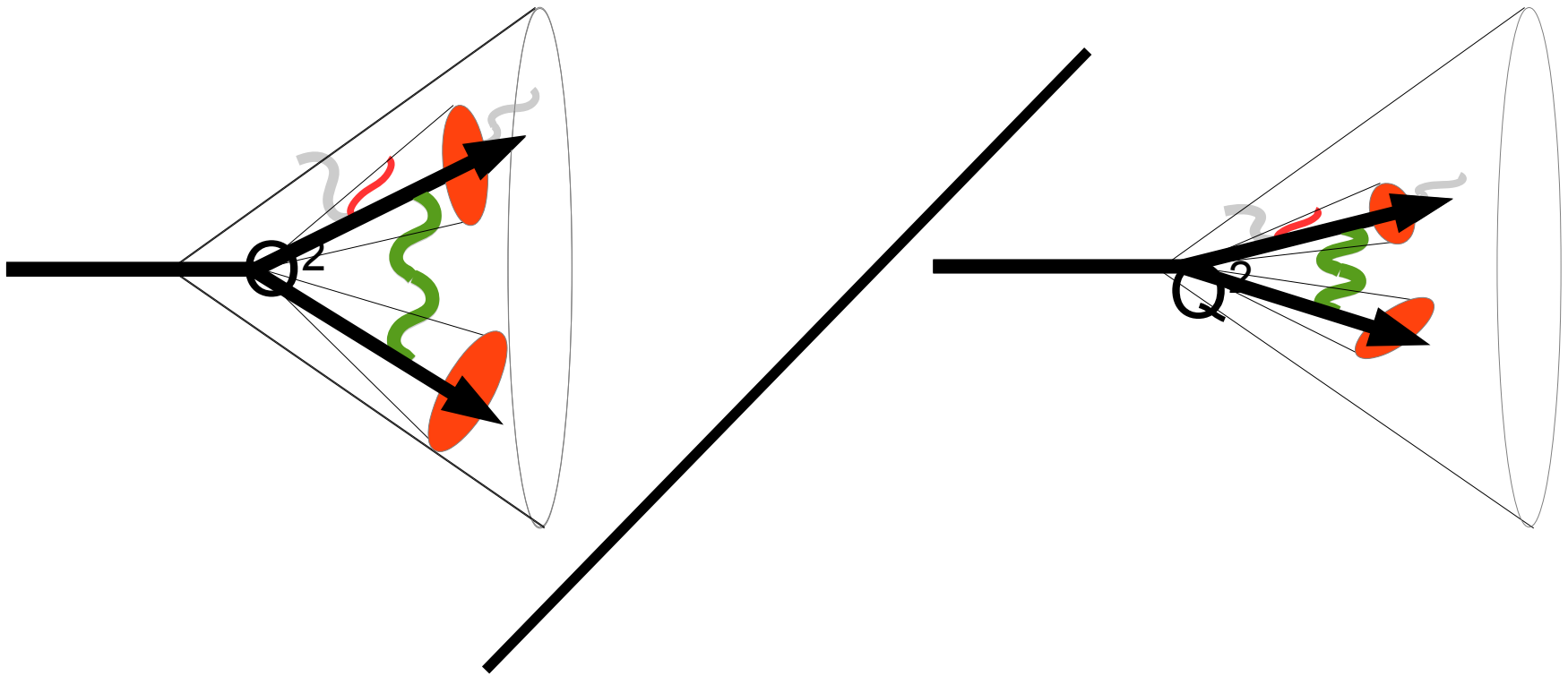


Jet Mass

$$M^2 \sim p_1 p_2 \Delta R_{12}^2$$

Tells us about the medium

Performing Measurements in HI



Many ideas to decorrelate jets against observables

Decorrelation allows for data based ratios

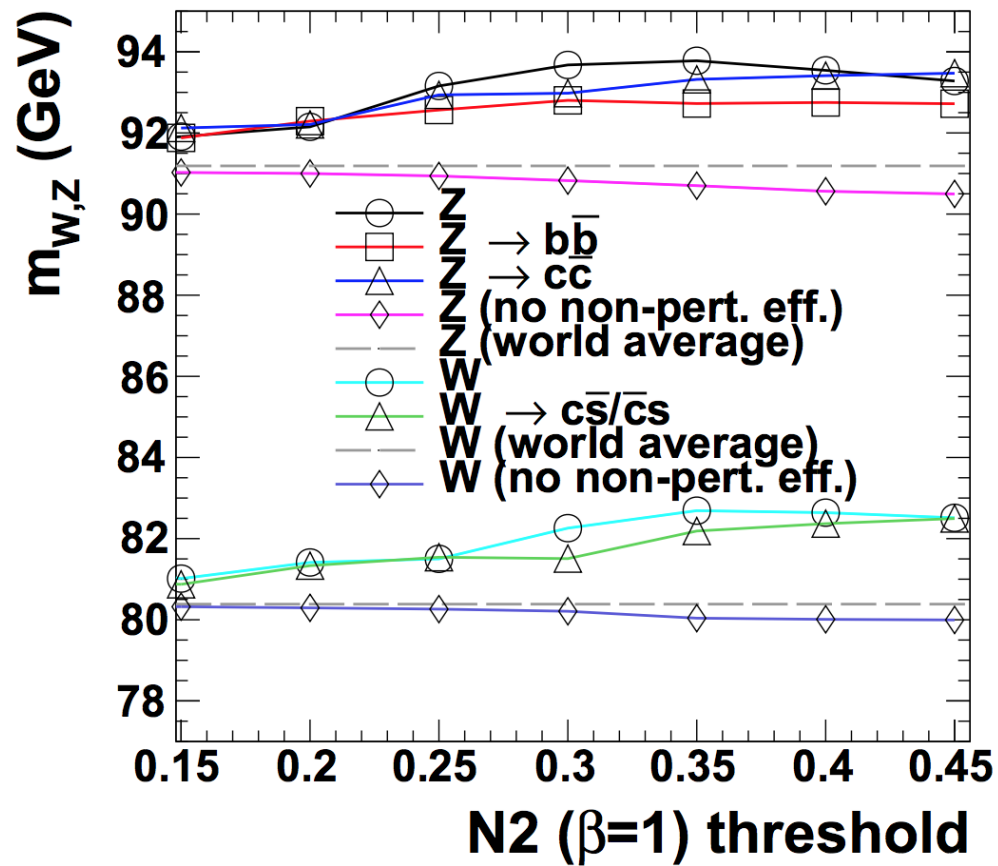
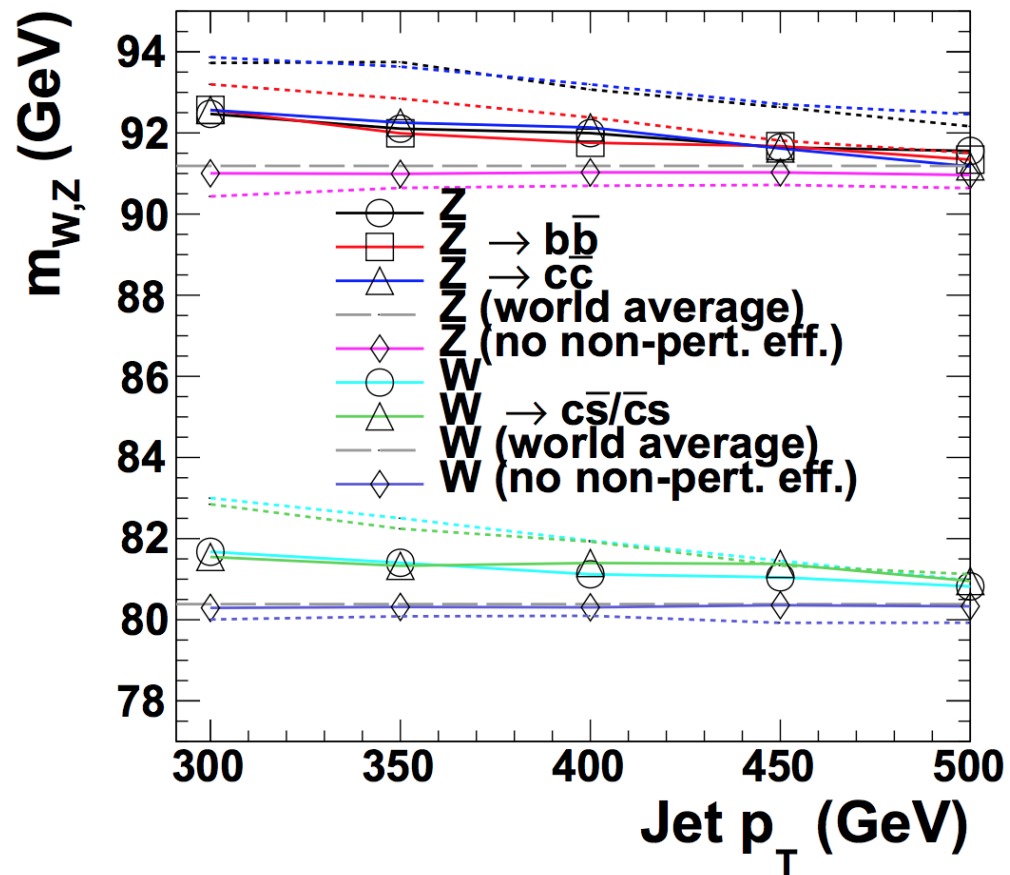
Allows us to be sensitive to small features

Summary

- The act of substructure is a tool that we can use
 - Allows us to sculpt jets to isolate specific physics
- For recent hadronic resonances
 - Constructed decorrelated observables against m_{SD} p_T
 - With ML we can decorrelate in higher dimensions
 - This can allow us to take care of systematics
 - Can lead to improved discrimination + decorrelation
- Extending beyond that....
 - Can we use decorrelation schemes to enhance HI?

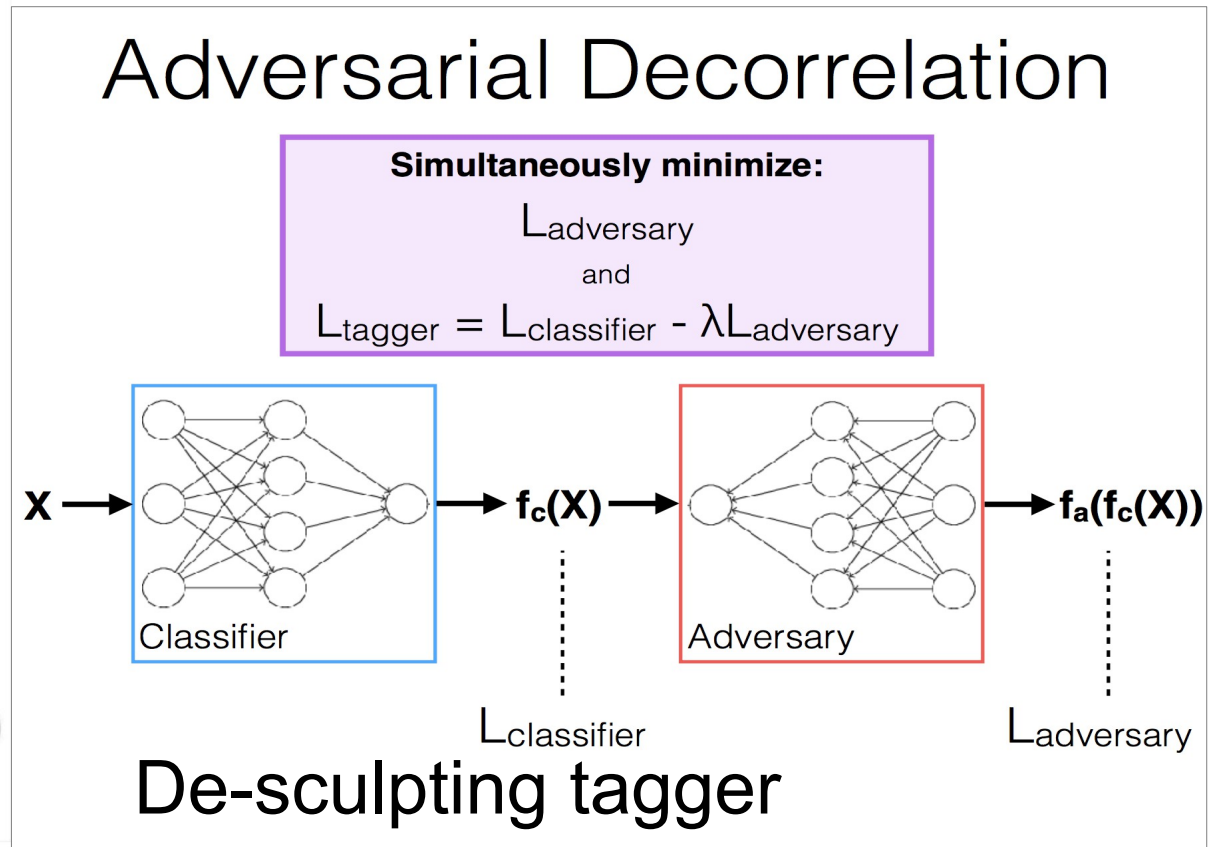
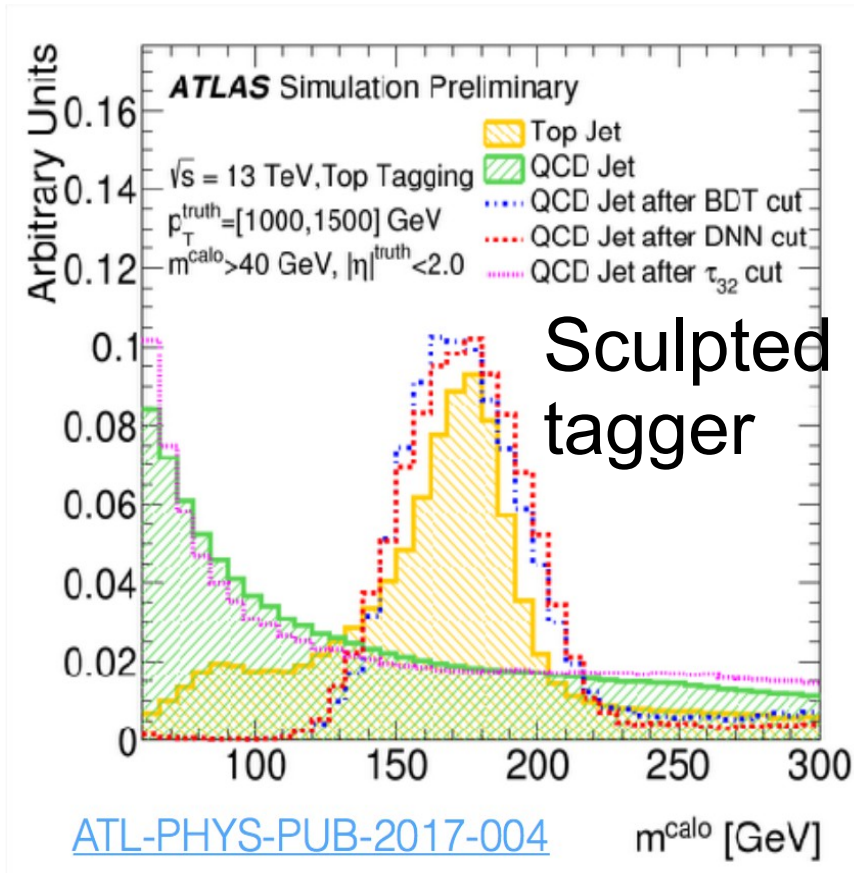


Thanks!



New Directions

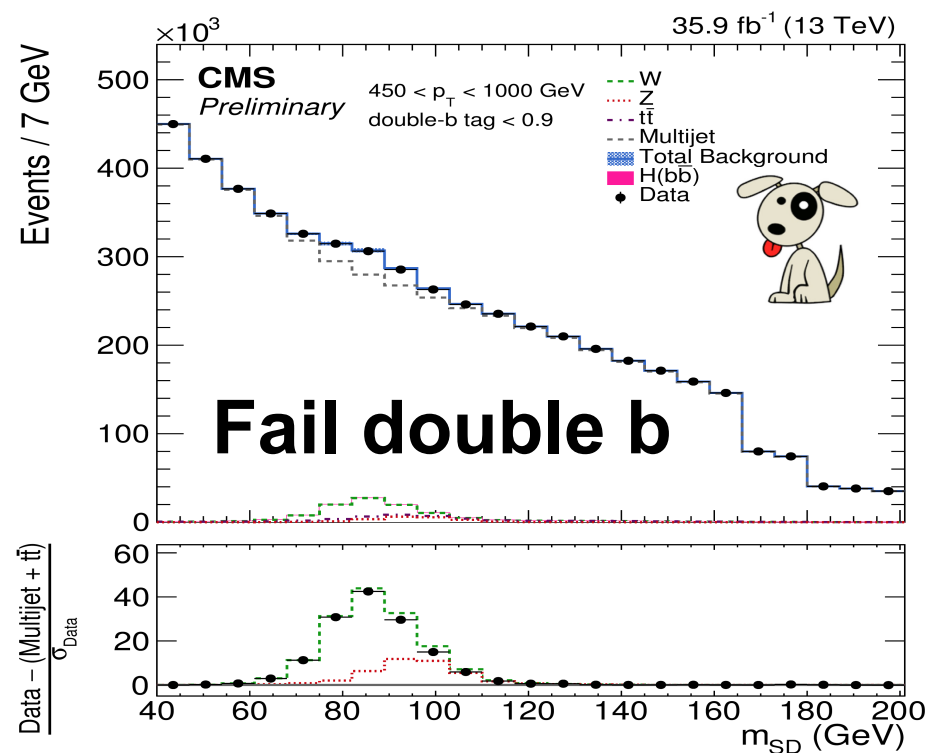
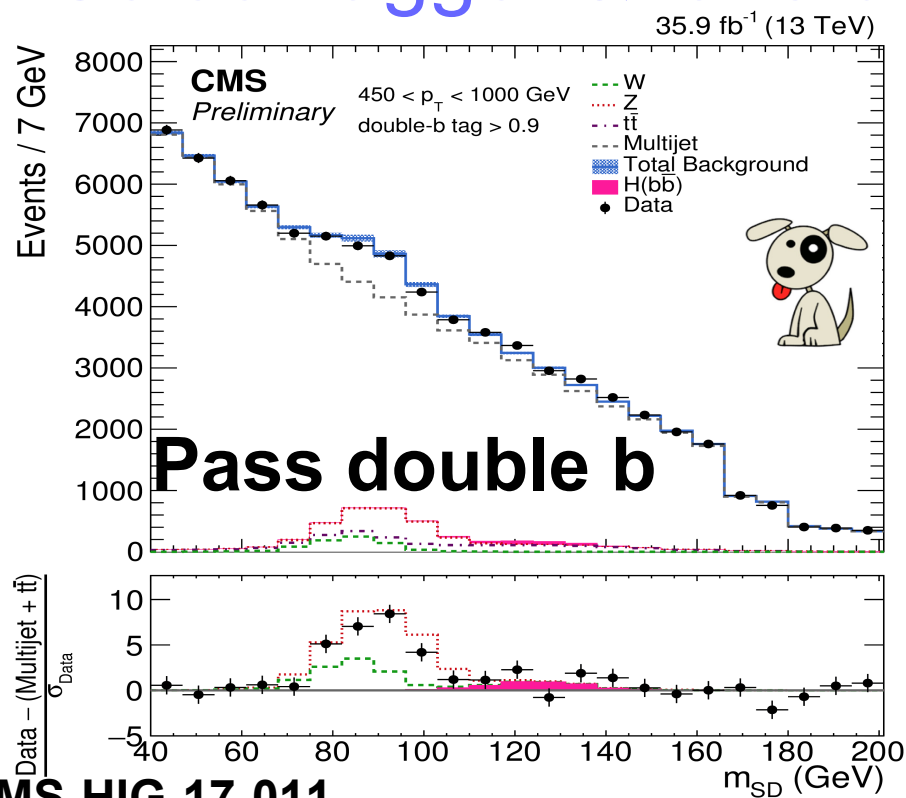
- Key concept was modifying jet substructure



New approaches w/deep learning **can decorrelate observables**

Ingredients to see Higgs

- Select a jet $p_T > 450$ GeV (single massive jet trigger)
- Require the jet to pass a 2 prong tag
- Require jet to be double b-tagged
- Select tagger & take both pass and failing selection



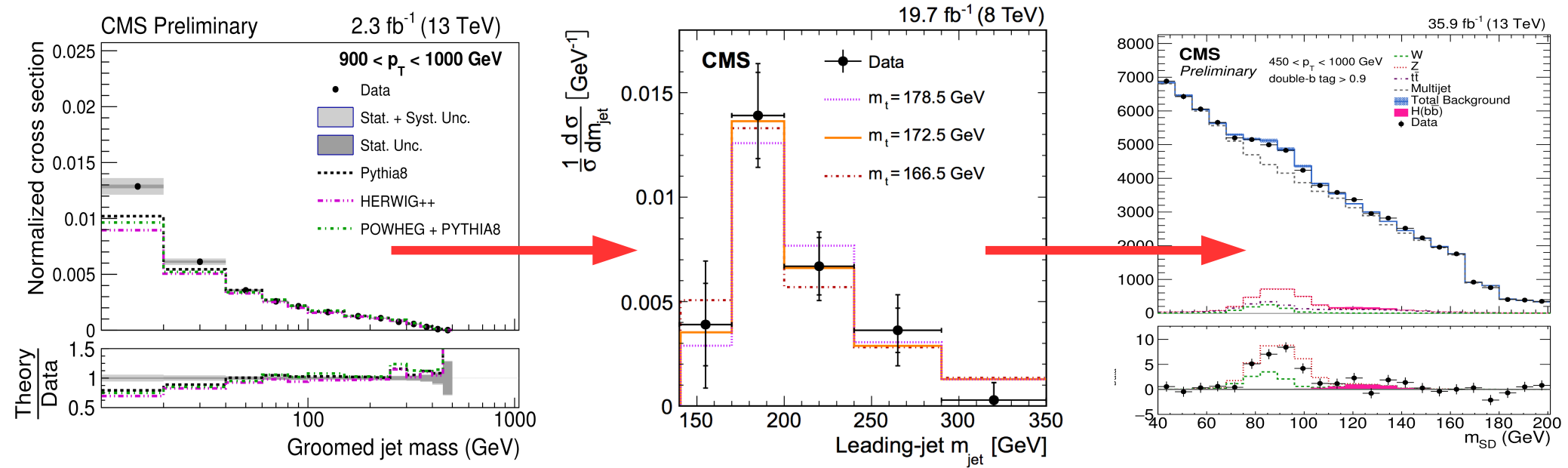
PDF

ME

Resum.

EWK

NP?



QCD meas. probing ever smaller event scale
 Going smaller build on previous measurements

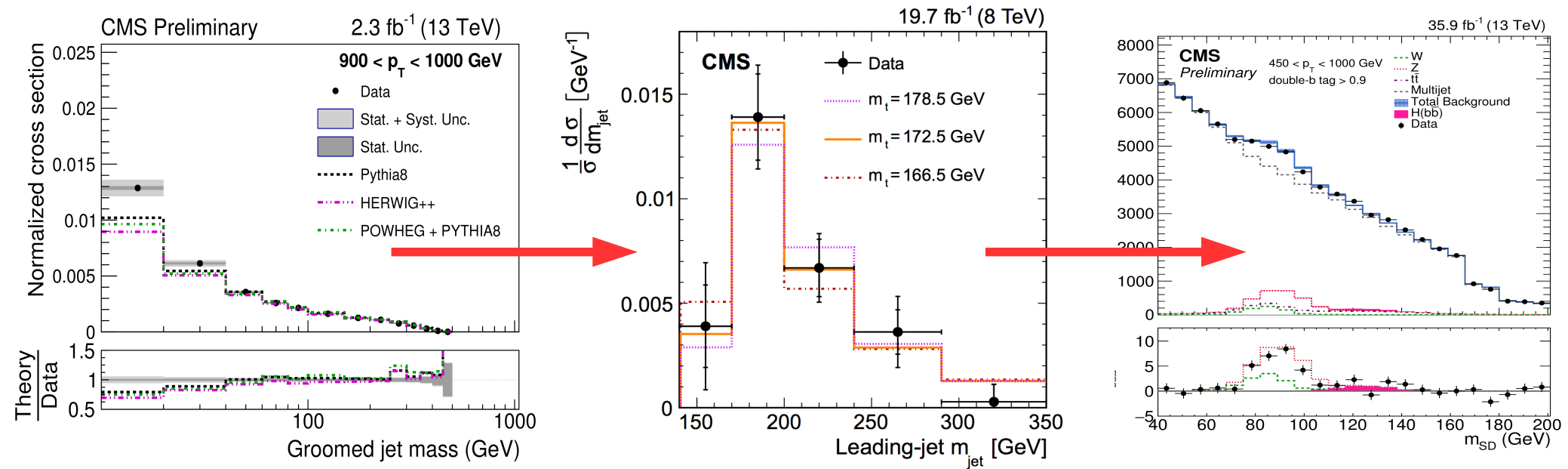
PDF

ME

Resum.

EWK

NP?



QCD meas. probing ever smaller event scale
 Going smaller build on previous measurements

A wealth of new and exciting results

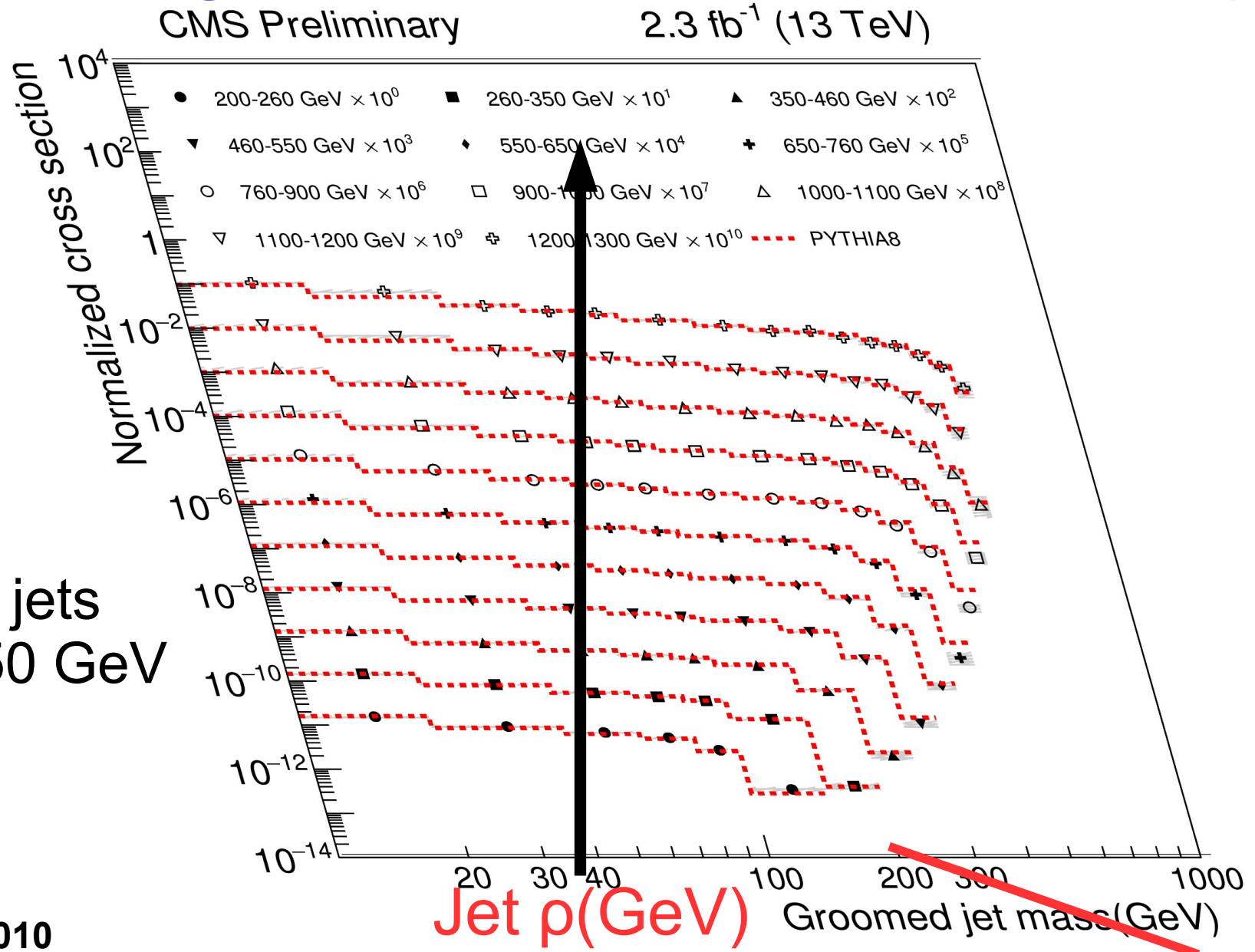
Standard model mass measurements

Probing the Quark/Gluon medium

High p_T Higgs

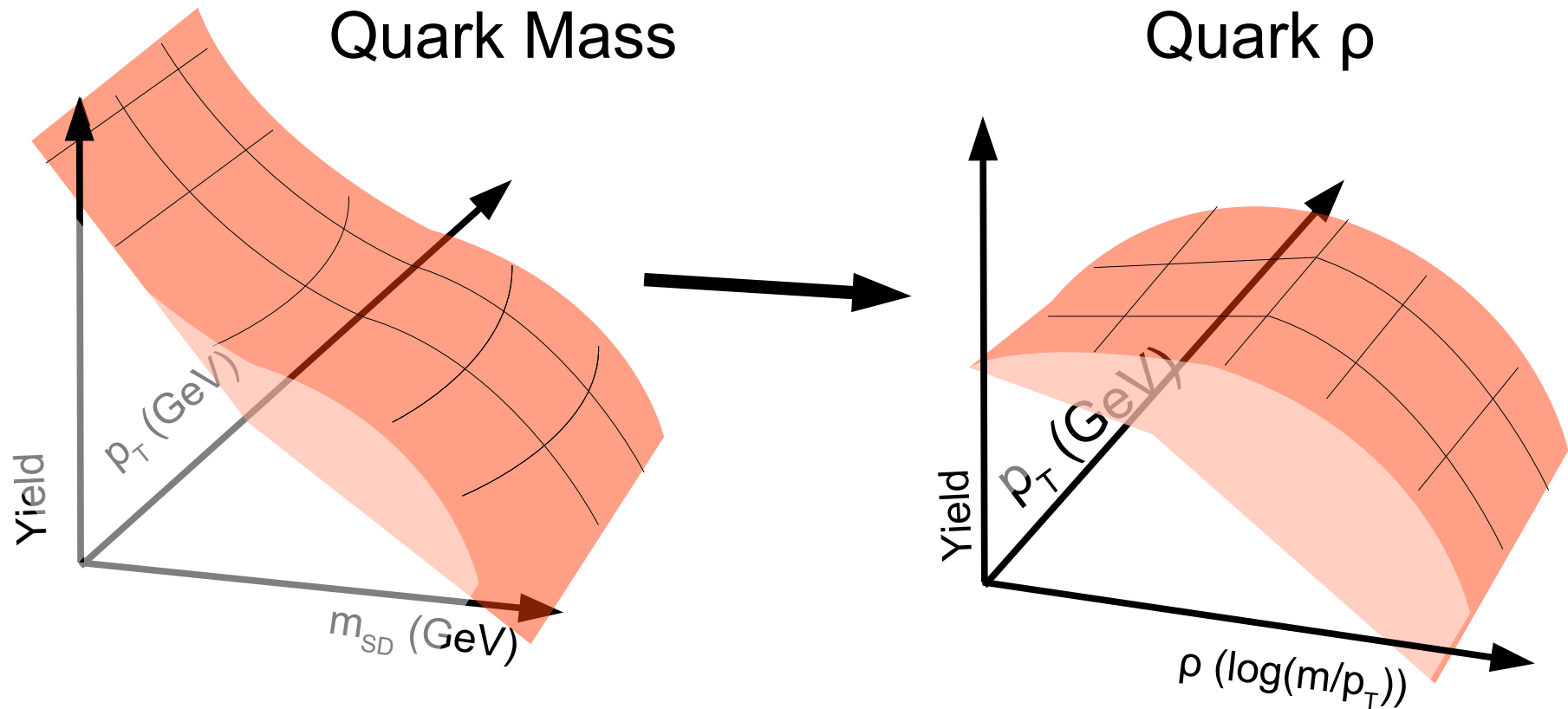
Groomed Evolution

When grooming the mass becomes even flatter in ρ



Merging with ρ

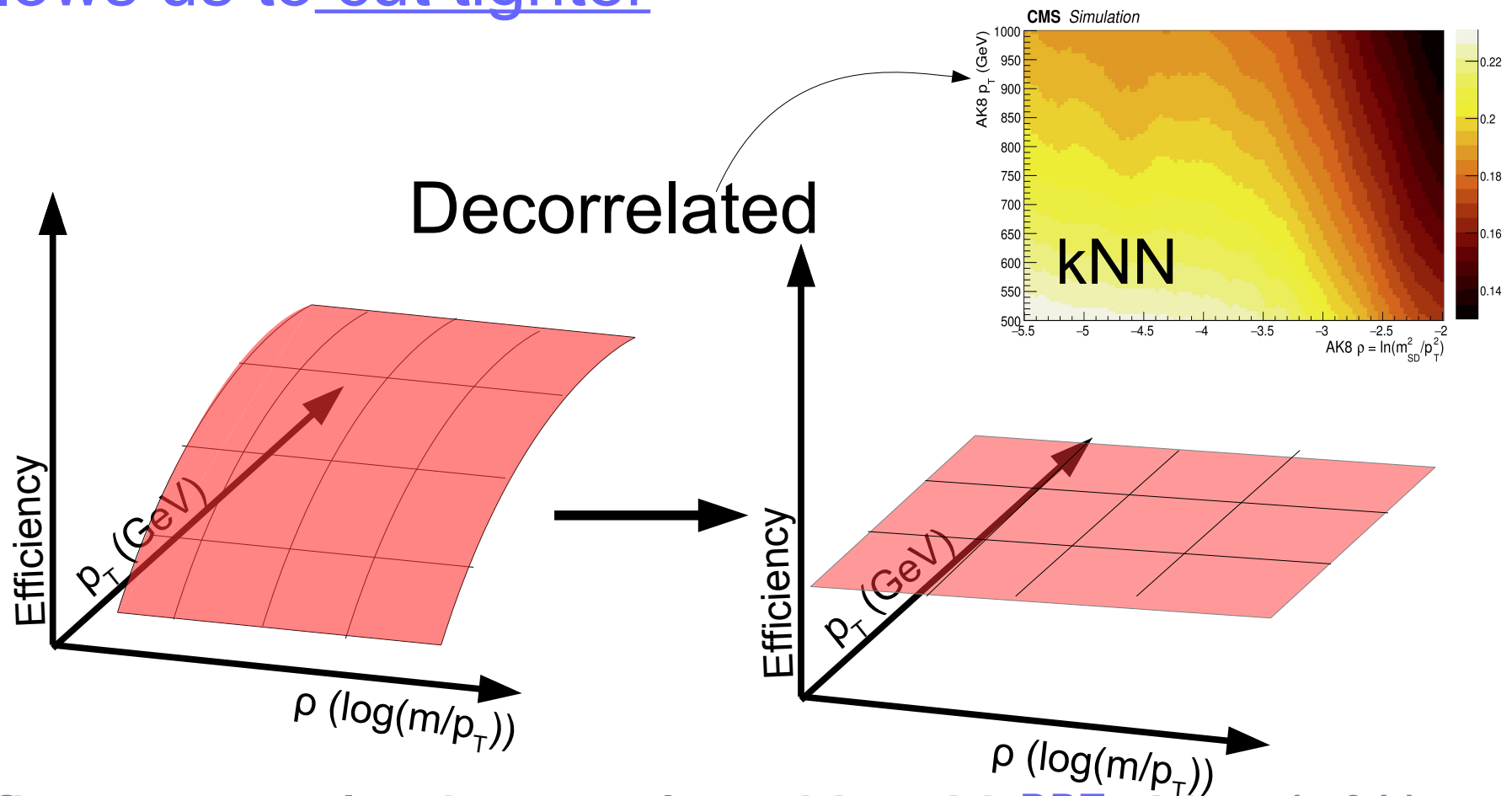
- Merging to ρ makes imposes invariance over p_T



When translating to ρ distributions over p_T are also invariant
 This allows us to extend our fit from 1D mass to 2D p_T and ρ

Design a transform to decorrelate against mass and p_T

Decorrelating avoid mass sculpting
allows us to cut tighter



As a first example decorrelate $N_2 \rightarrow N_2^{DDT} = N_2 - \epsilon(x\%)$

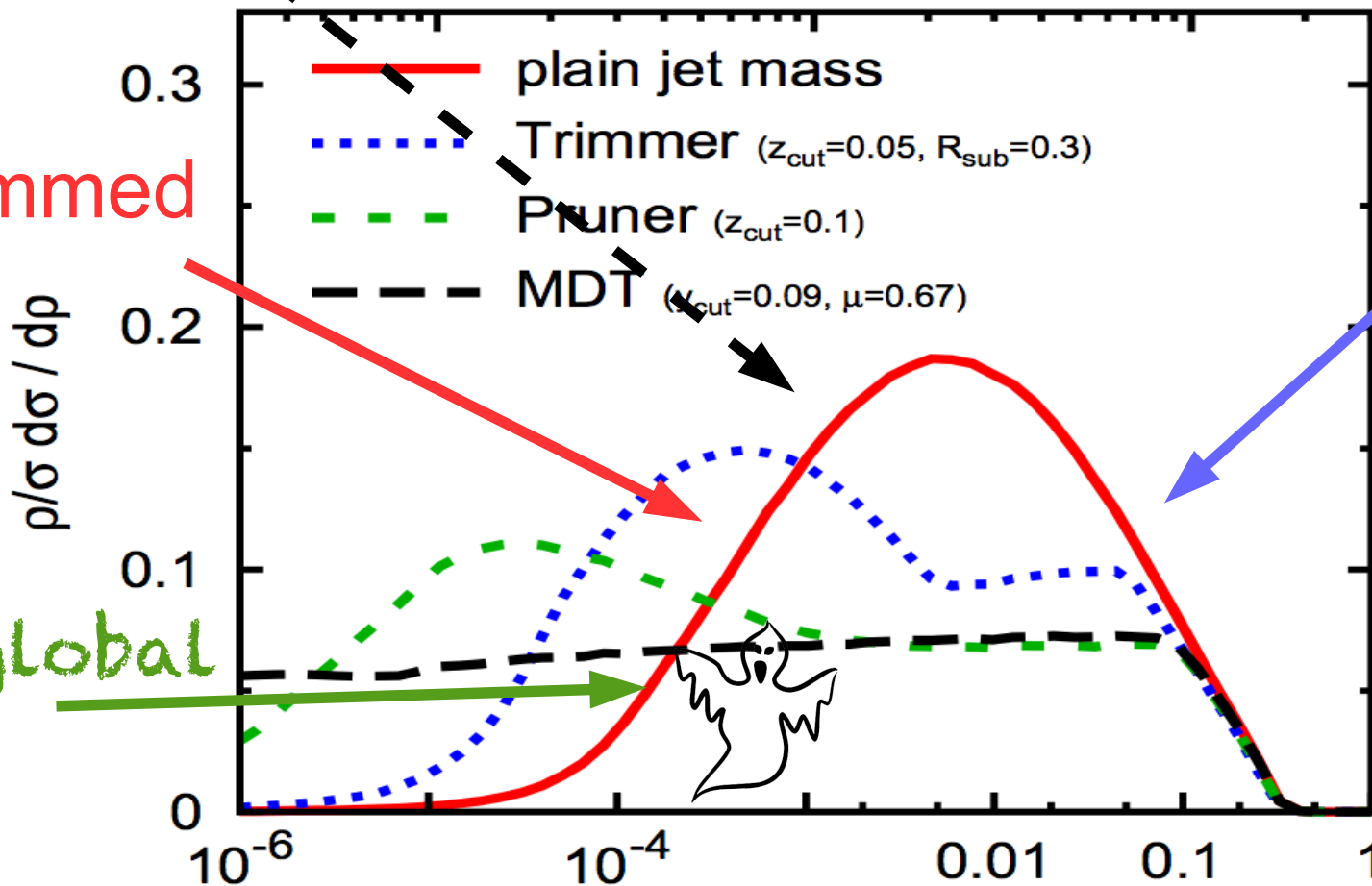
Anatomy of Jet Mass: Theory

Sudakov Peak

quark jets (Pythia 6 MC)

m [GeV], for $p_t = 3$ TeV, $R=1$

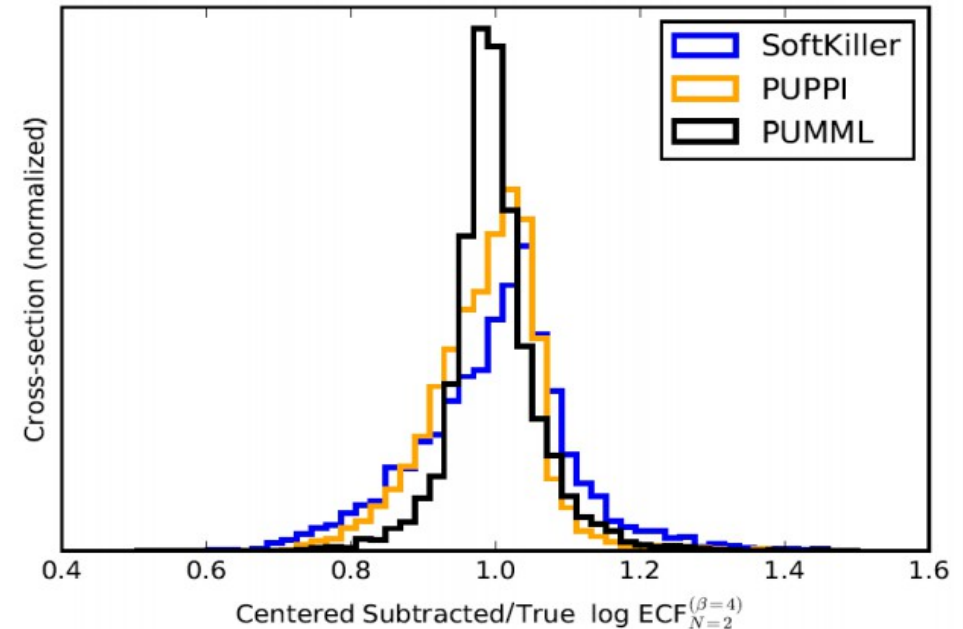
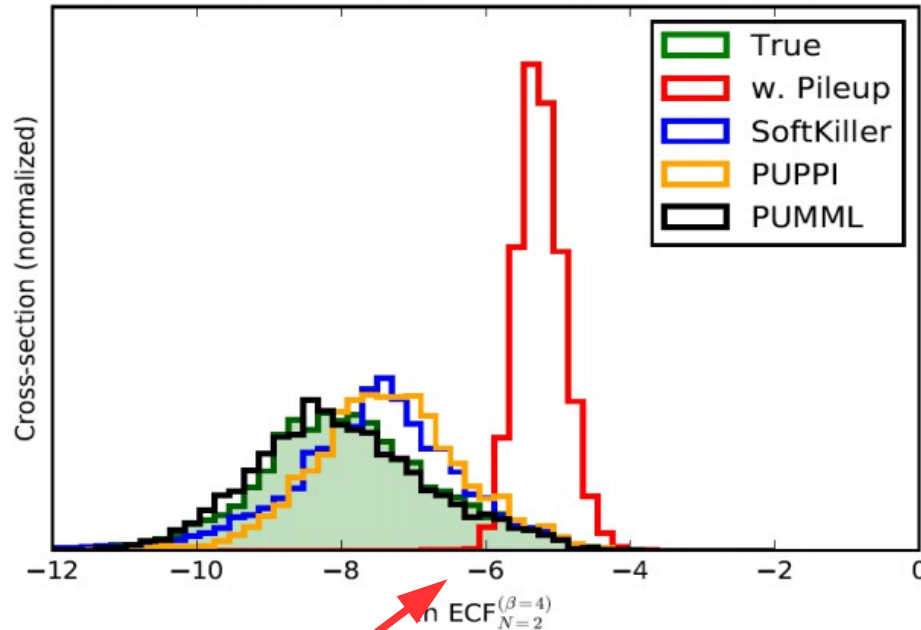
10 100 1000



Calculate in ρ invariant over mass/ p_t

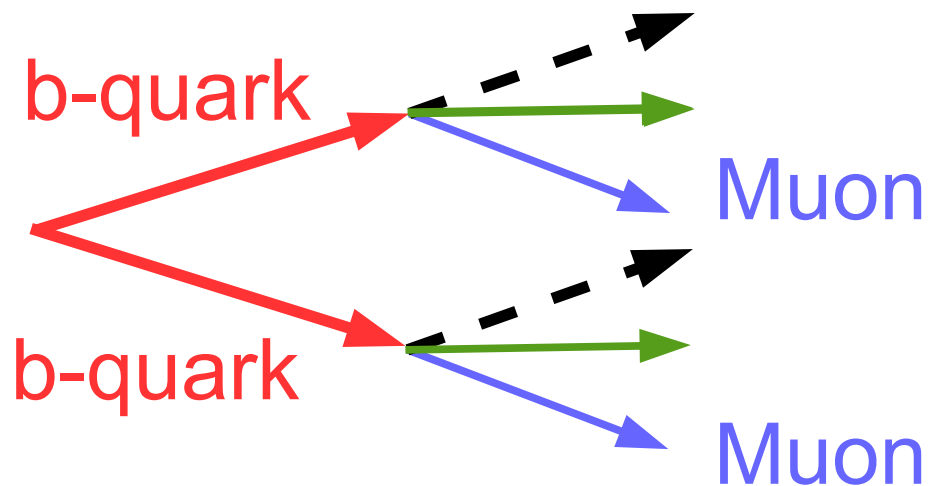
New directions II: PUMML

PU mitigation w/Machine Learning

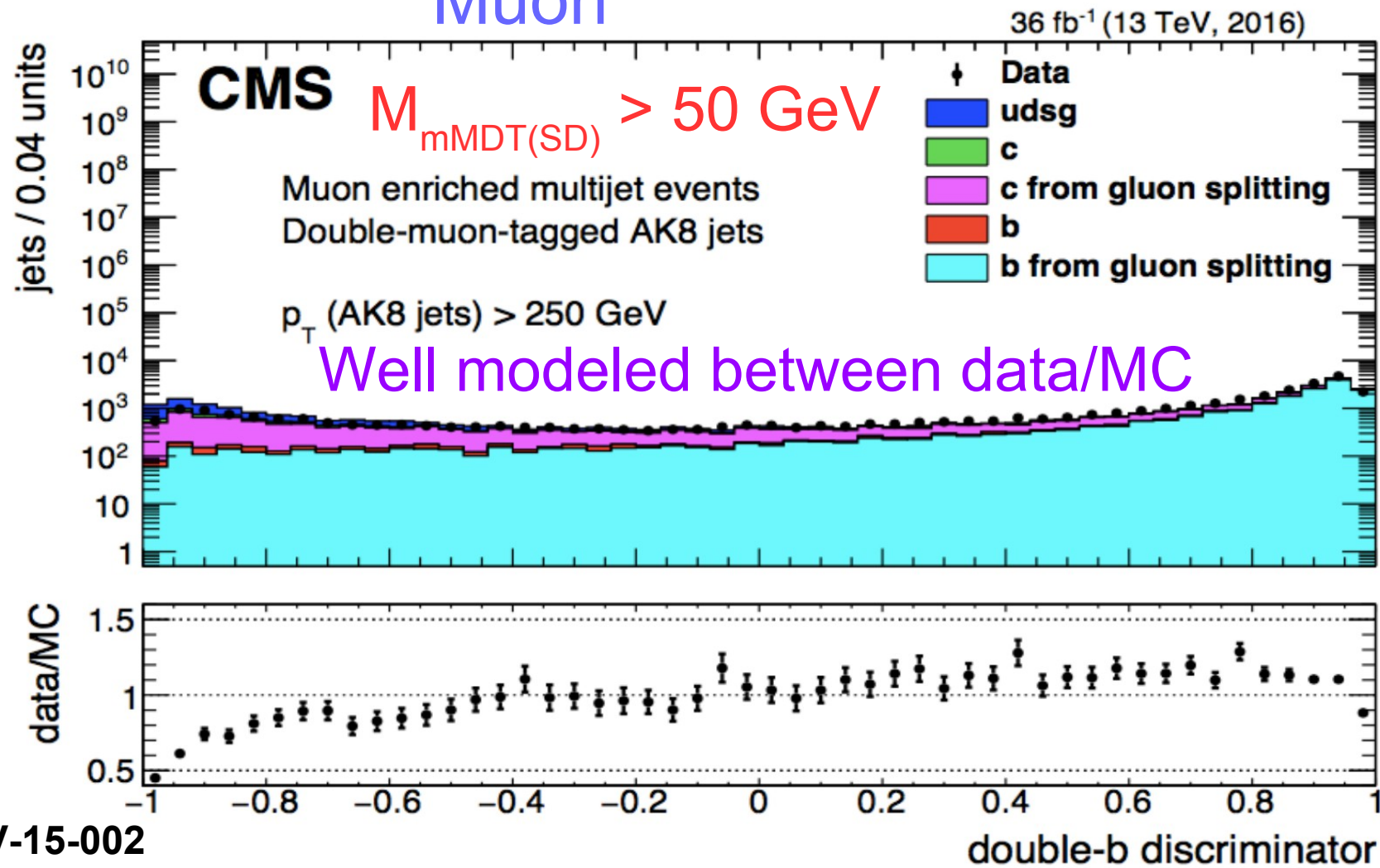


Understanding jets requires resolving jet properties in the **dense environment of the LHC** (known as pileup)

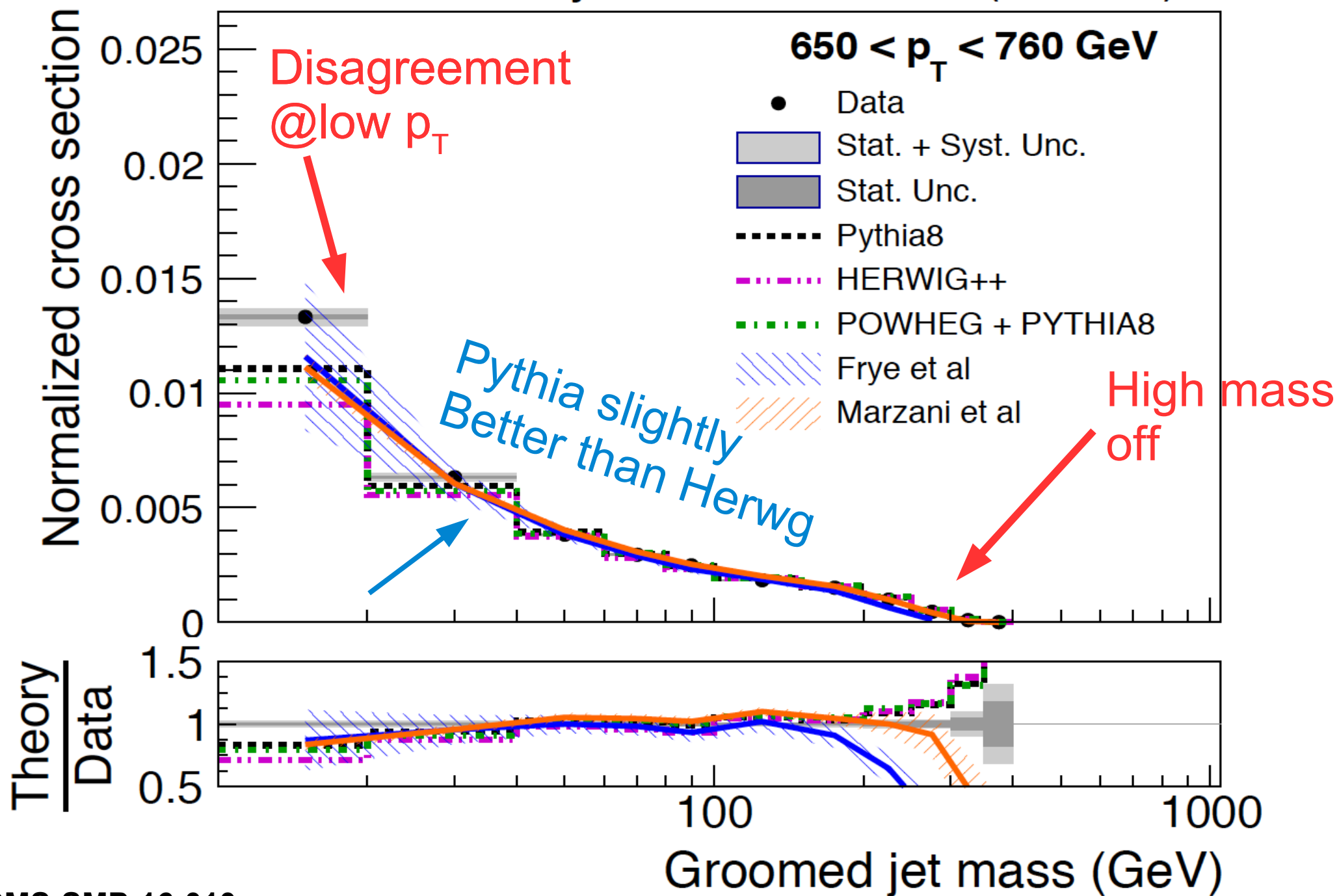
What about data/MC



Tag two muons in a jet
Use this to infer signal-like
2 b-quarks in a jet



CMS Preliminary

 2.3 fb^{-1} (13 TeV)

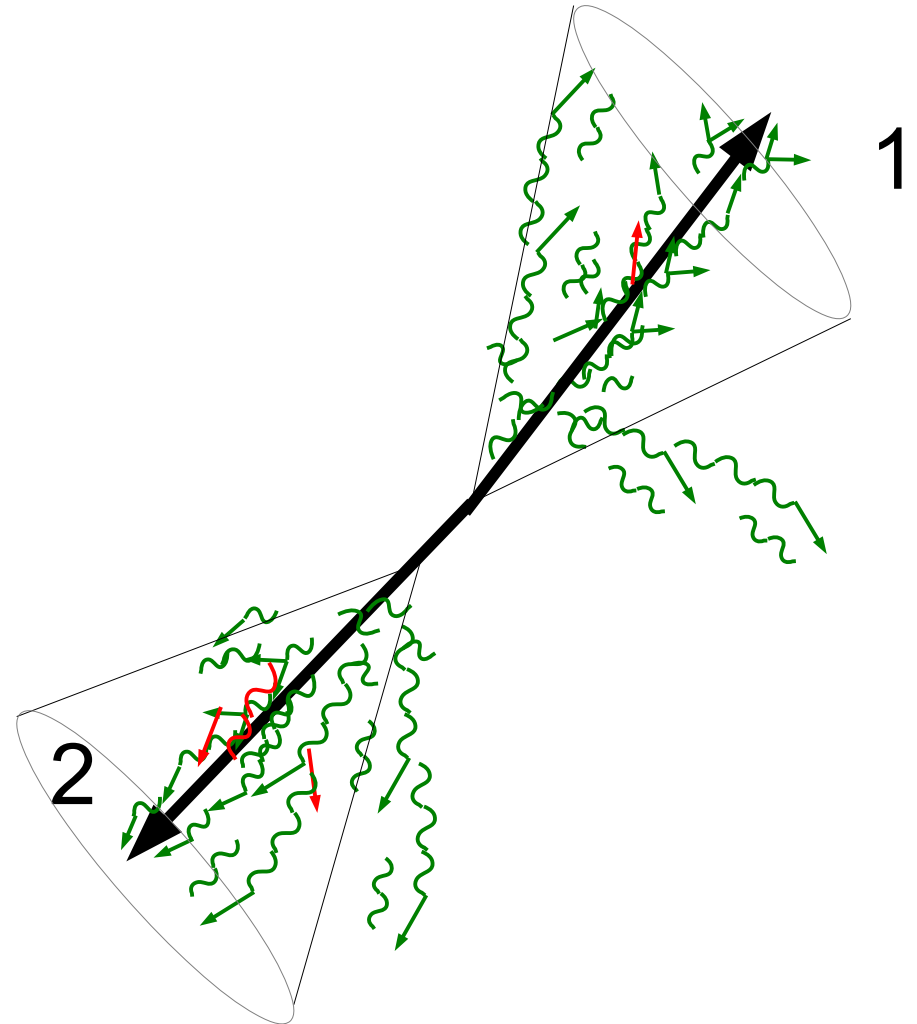
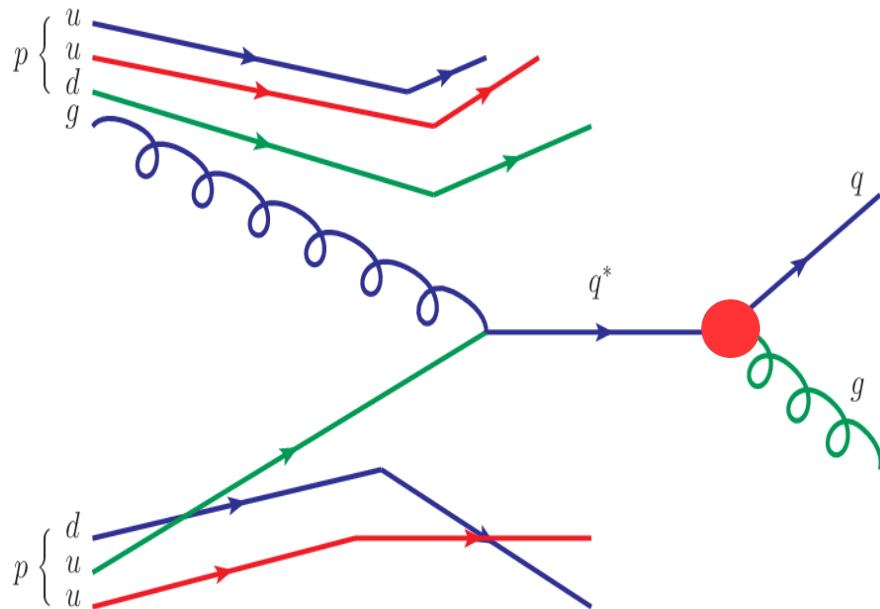
Basic Jet Measurements

In olden days :

Cared about the dijet process

The concern here :

good jet reconstruction to
resolve partons



Collinear splittings

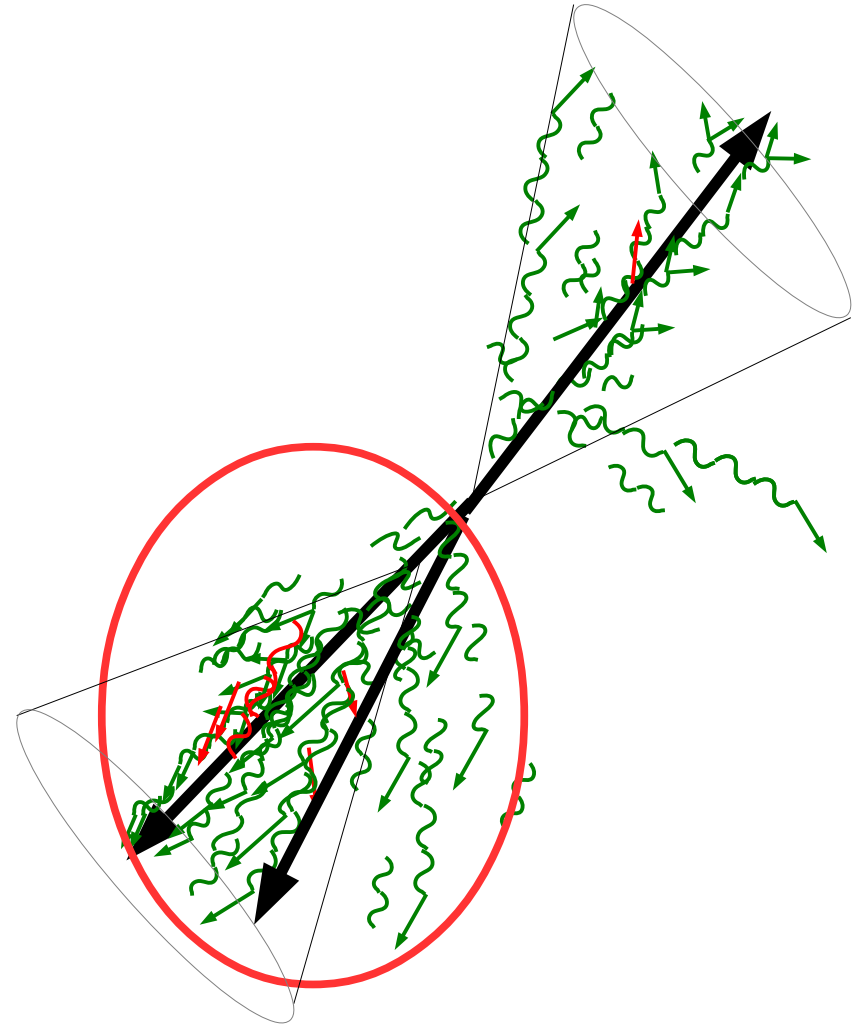
Currently:

Care about collinear splittings

The concern here :

Jet mass is a fundamental
test of QCD

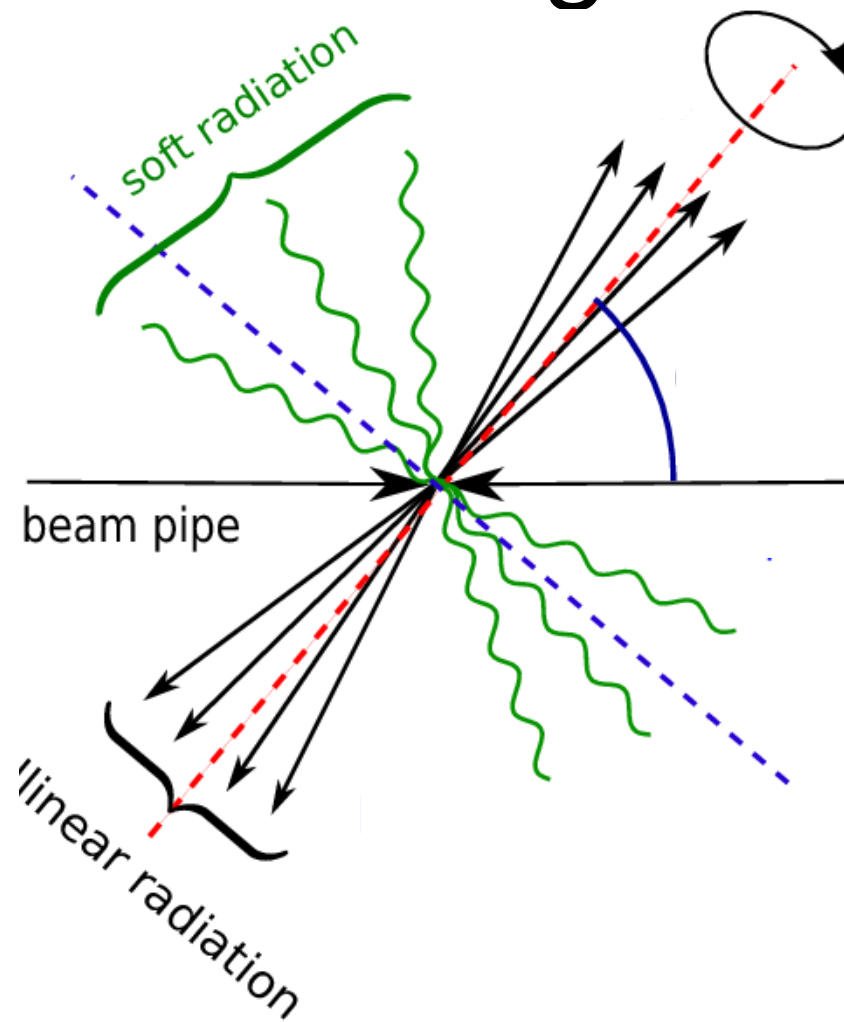
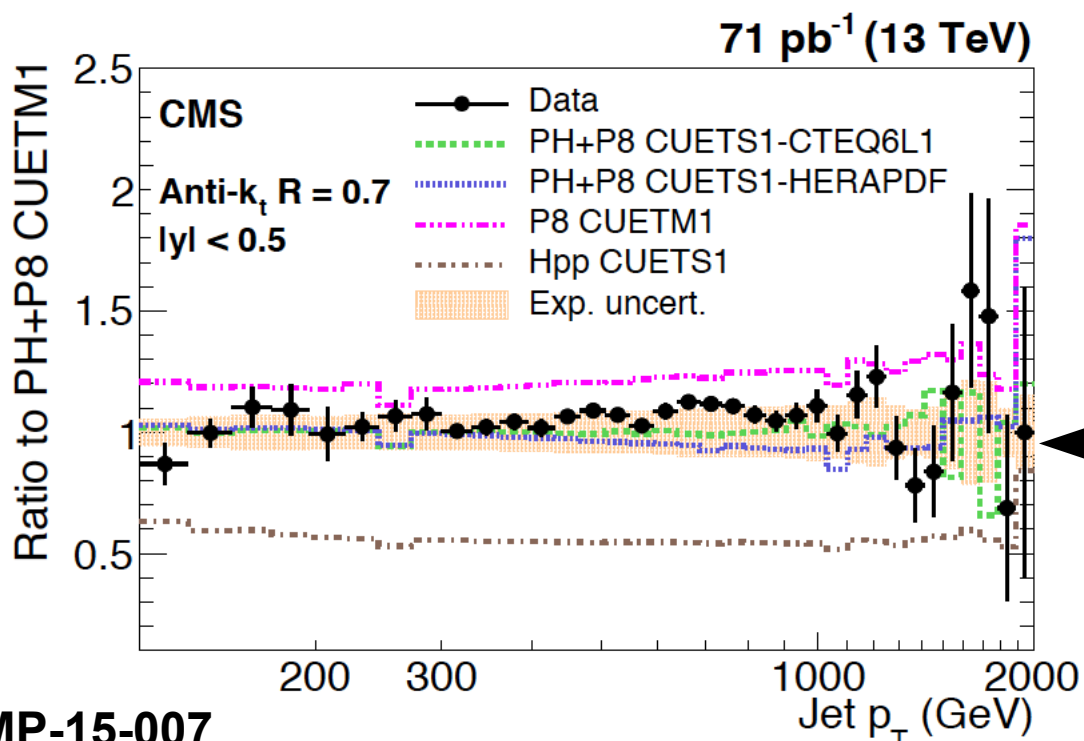
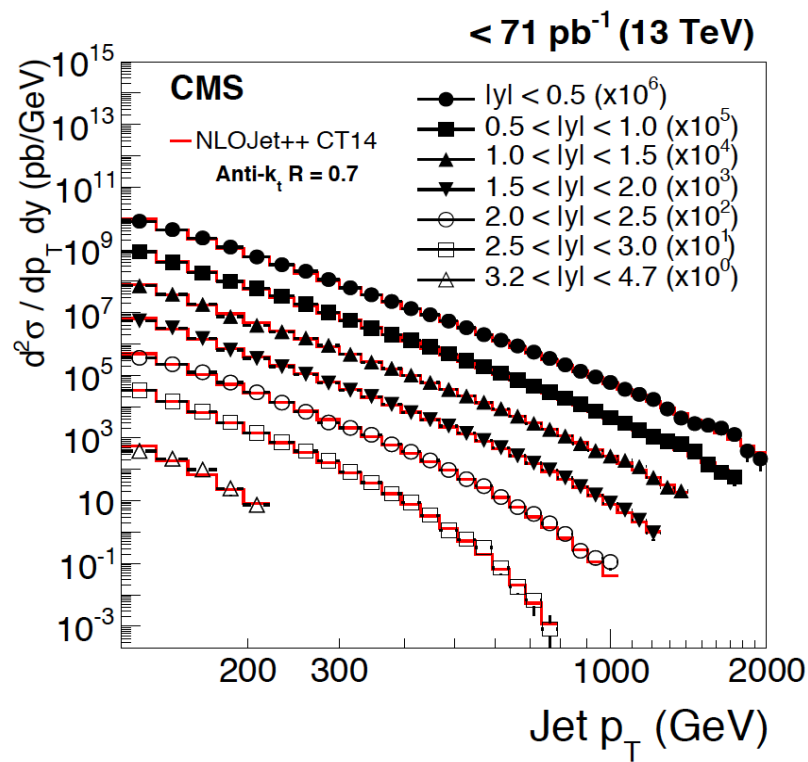
Also a tool for more?



In the case about the individual process of the jet

This will be focus of the rest of this talk

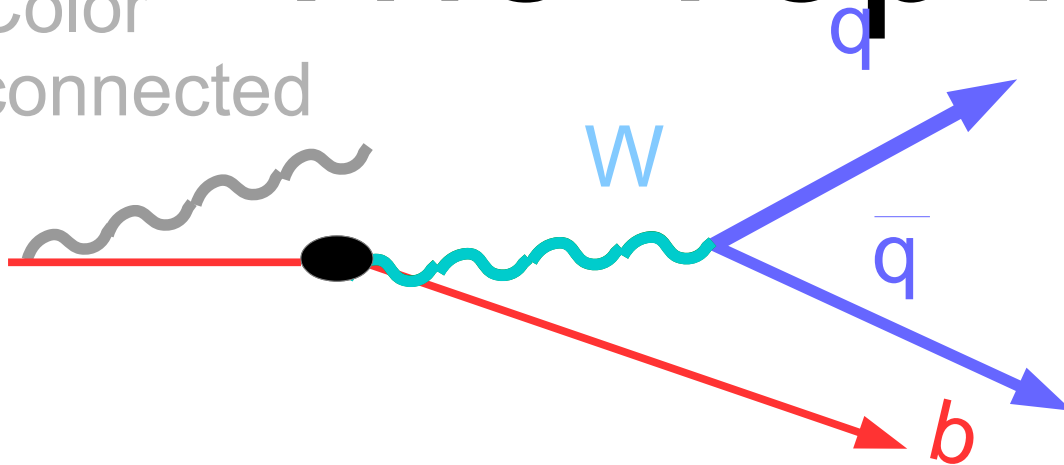
Understanding them



Basic jet
measurements
Allow us to **probe**
the production

The Top Mass

Color
connected

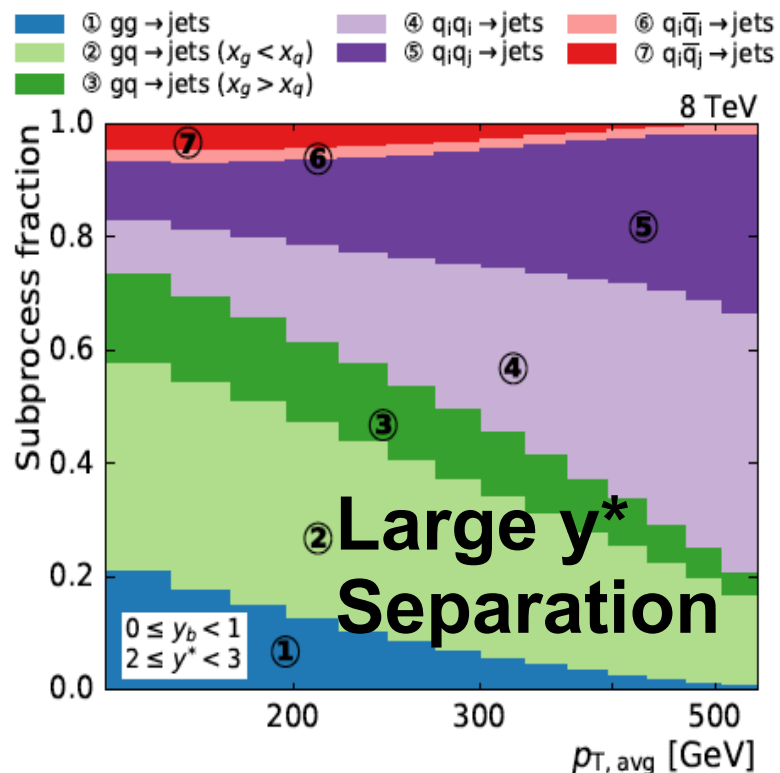


Top decays are a great probe of complicated objects

Complicated object with rich topology

Mass measurement requires understanding them

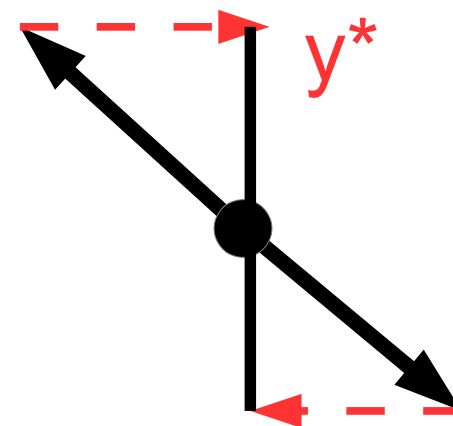
Understanding them



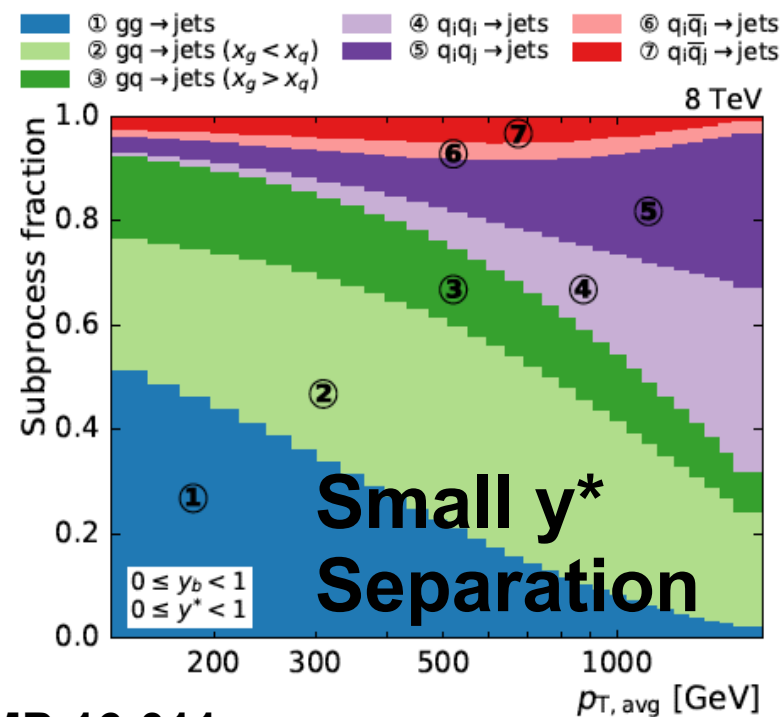
Quark
initiated

$$y_b = |y_1 + y_2|/2$$

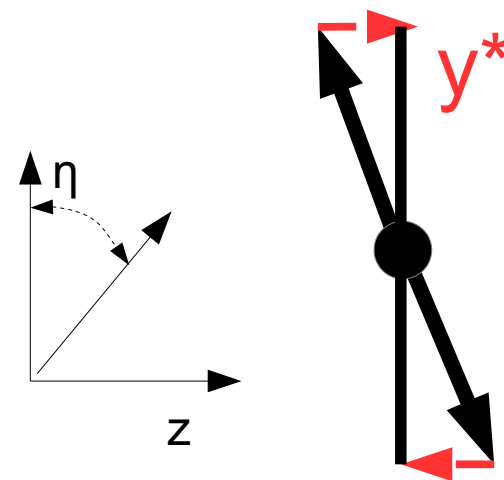
$$y^* = |y_1 - y_2|/2$$



Binning the dijet
kinematics allows
for constraints on
different diagrams



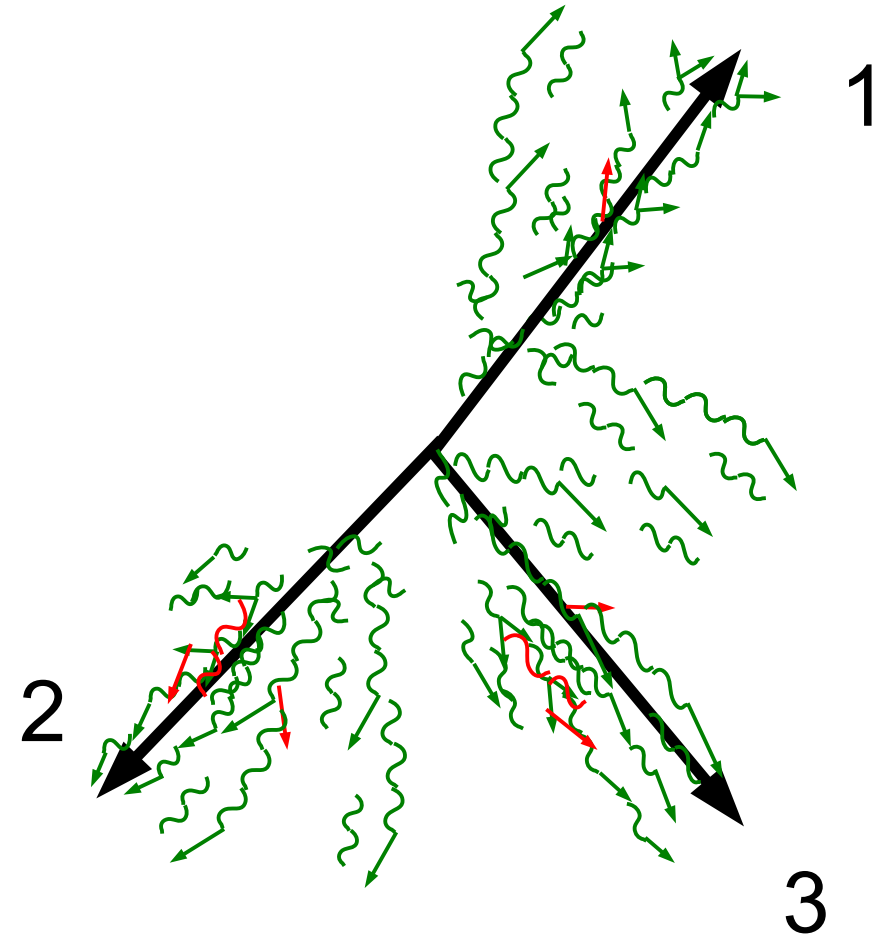
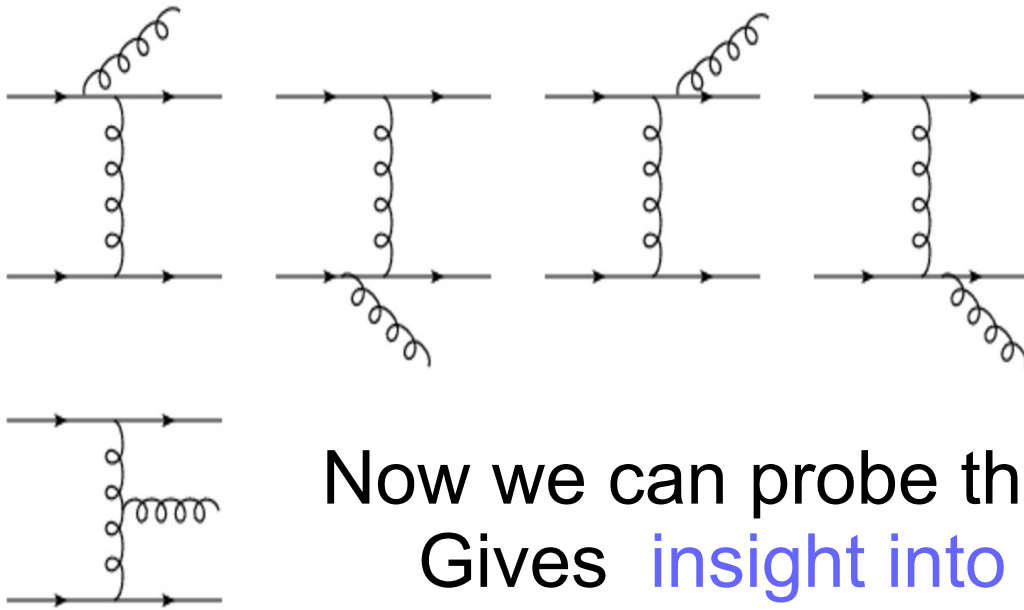
Gluon
Initiated



Basic Jet Measurements

In slightly less old days:
Cared about 3 jet process

The concern here :
Split out partons by type and
study additional QCD
properties

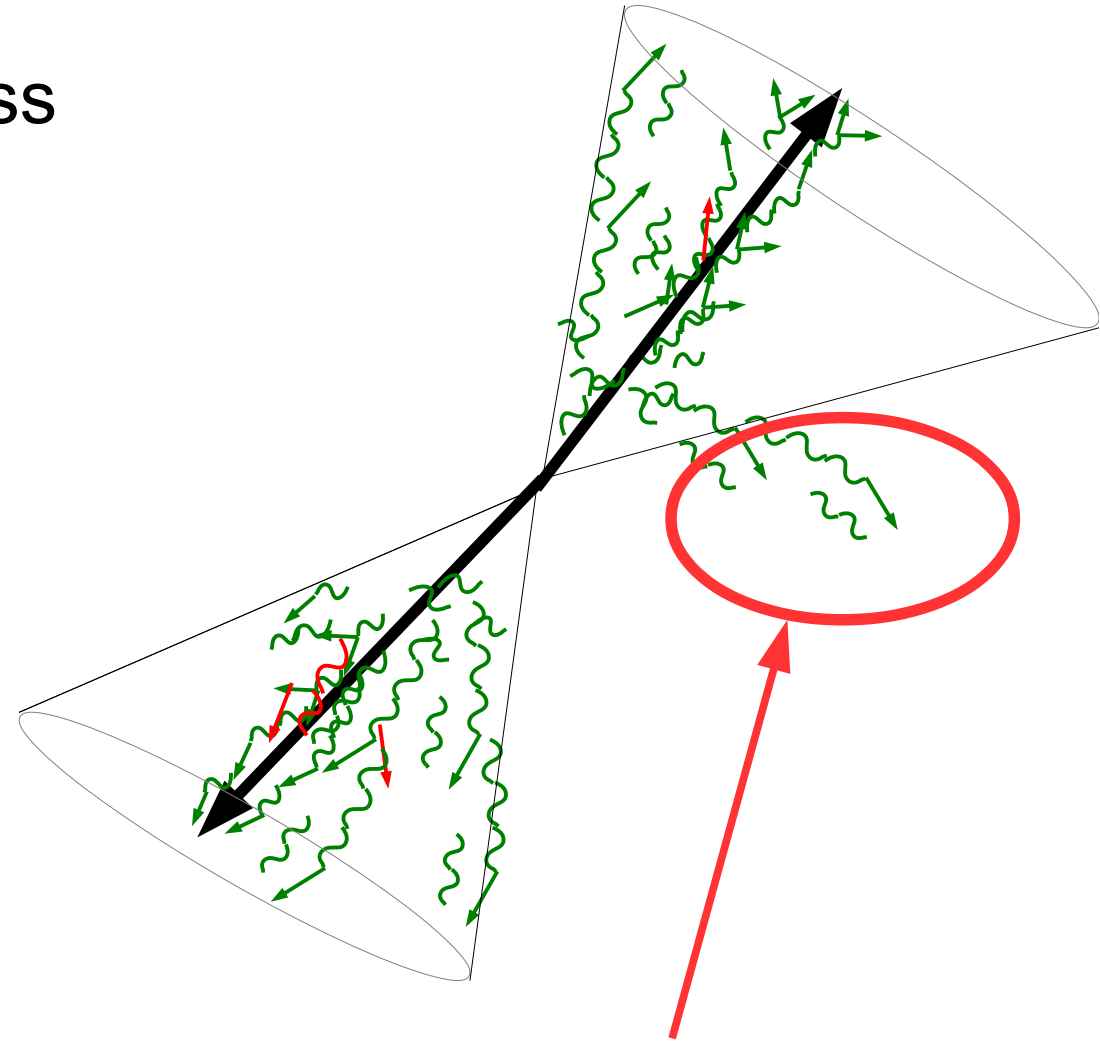
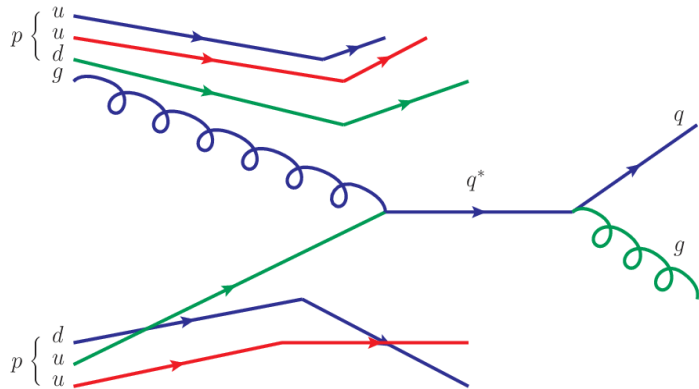


Now we can probe the 3rd radiated jet :
Gives insight into ME splittings

Basic Jet Measurements

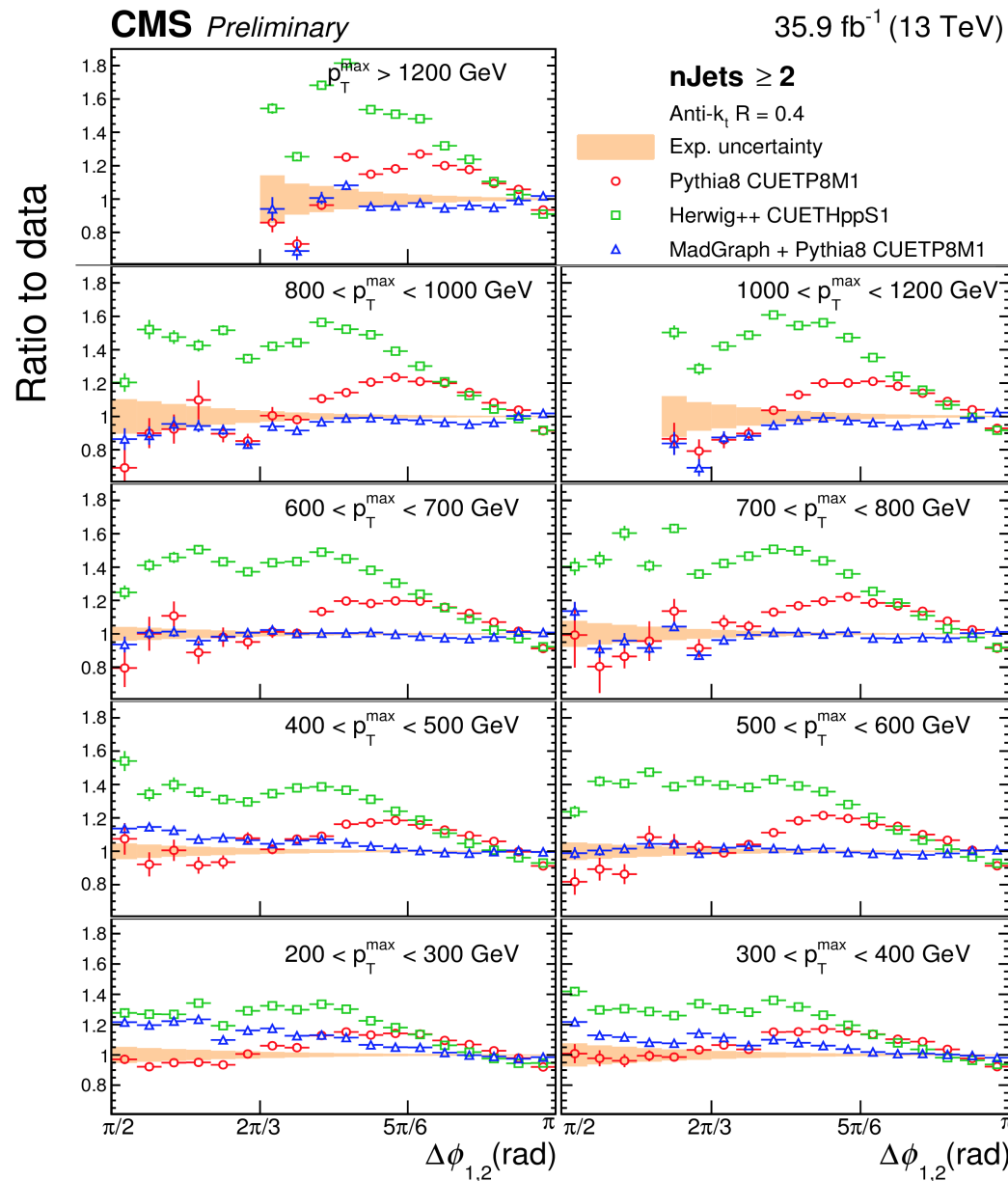
In olden days :
Cared about the dijet process

The concern here :
good jet reconstruction to
resolve partons



Aim of the game is to capture
the full parton

2 Jet Measurement



Pythia
2 jet

Pythia

Herwig
2 jet

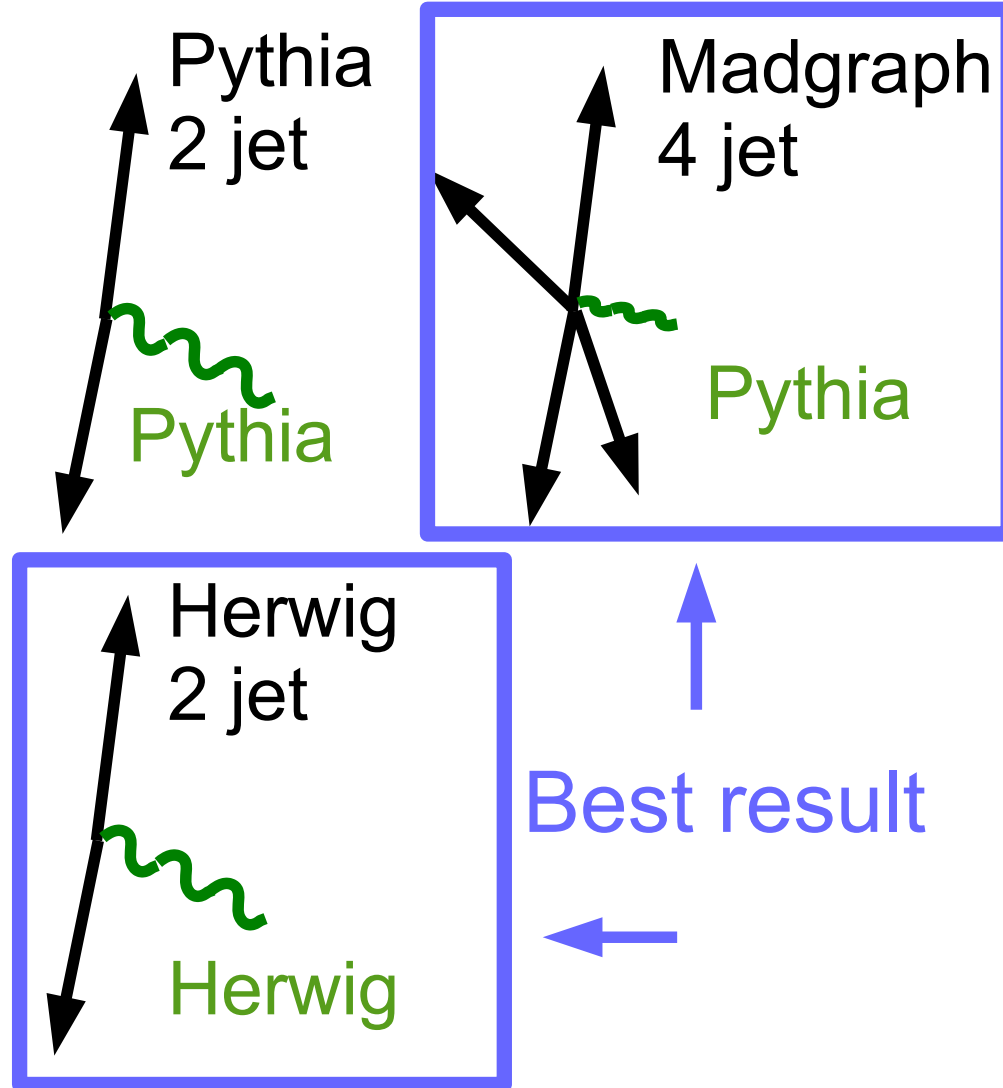
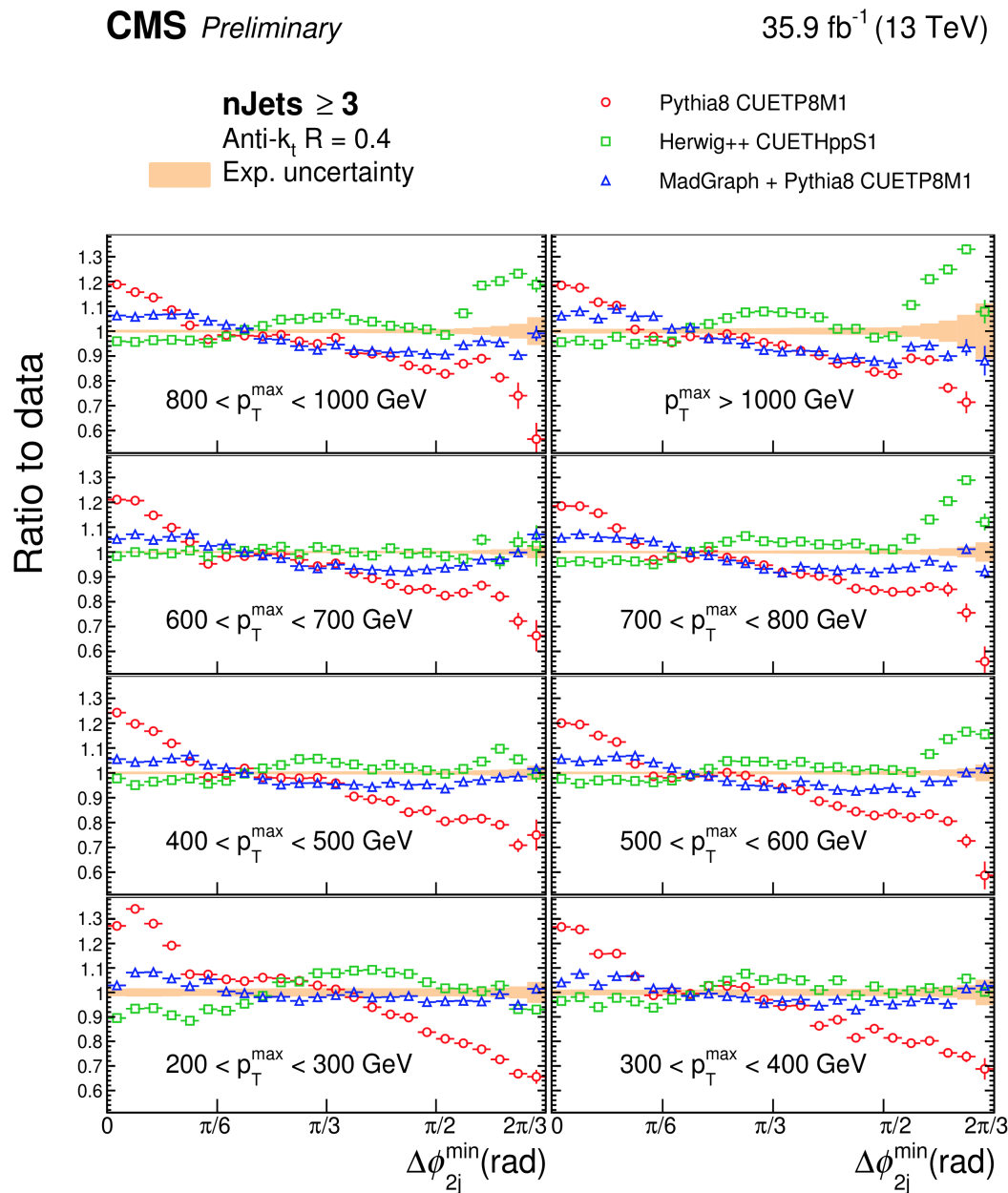
Herwig

Madgraph
4 jet

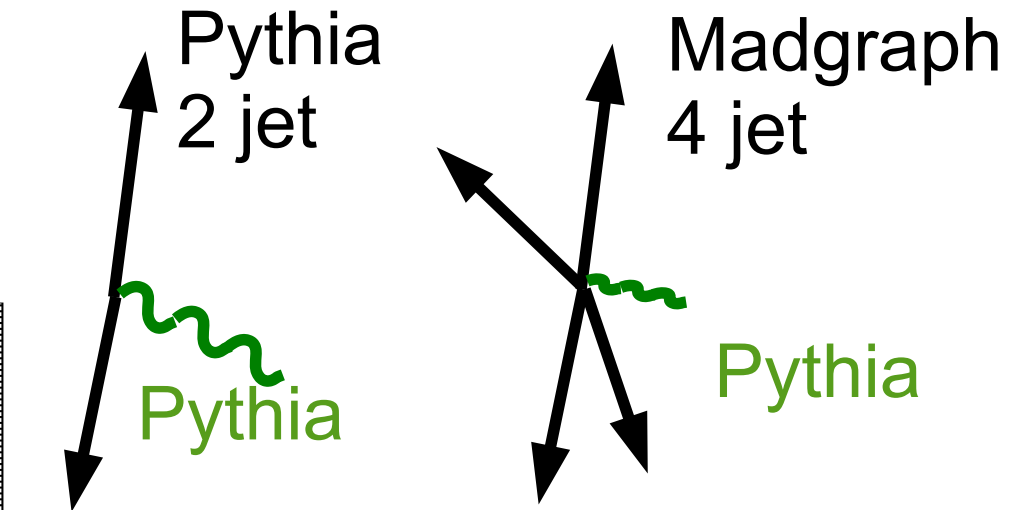
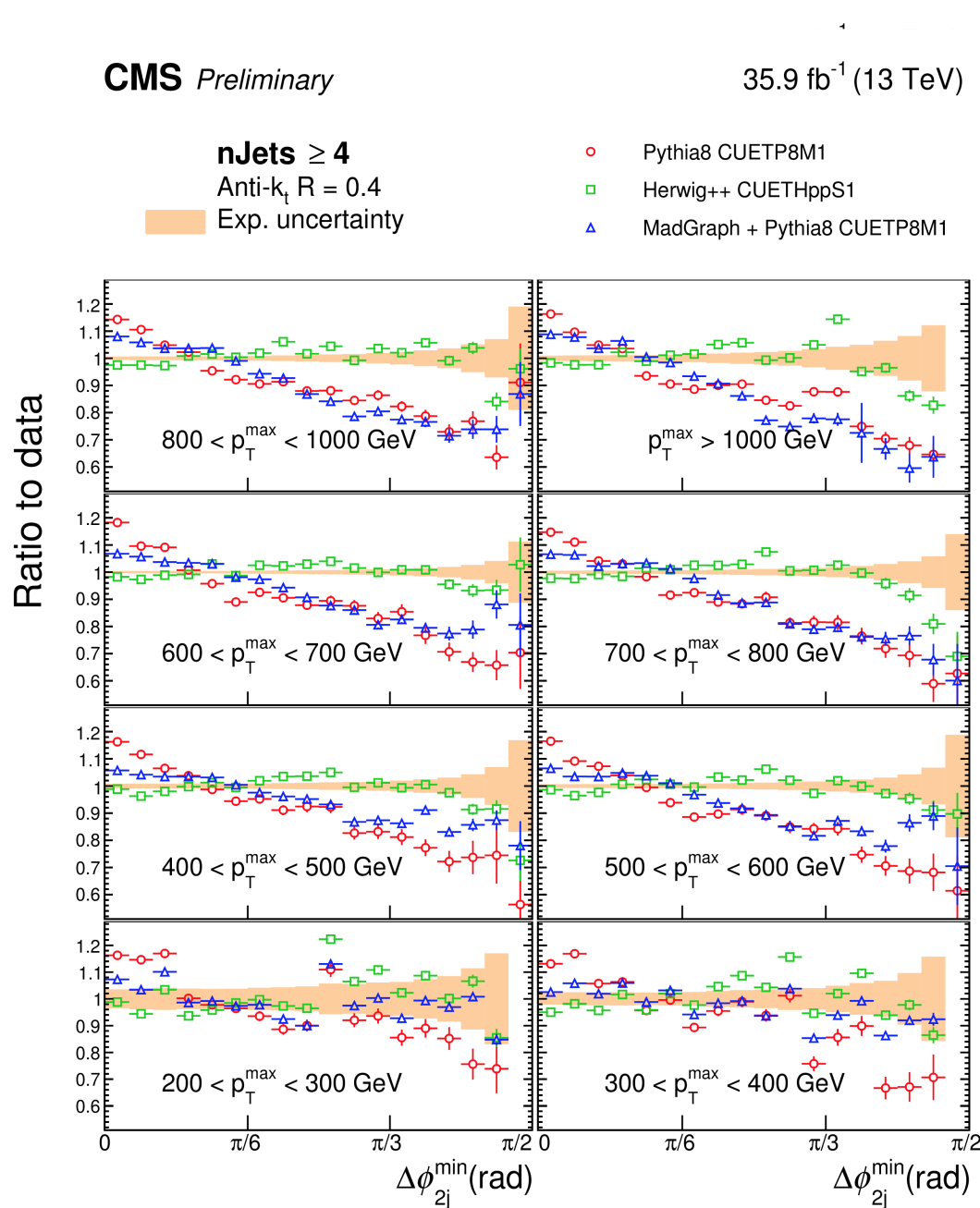
Pythia

Best result

3 Jet Measurement



4 Jet Measurement



Observation:
 Herwig models parton shower
 ME level an improvement

Collinear splittings

Currently:

Care about collinear splittings

The concern here :

Jet mass is a fundamental
test of QCD

Also a tool for more?

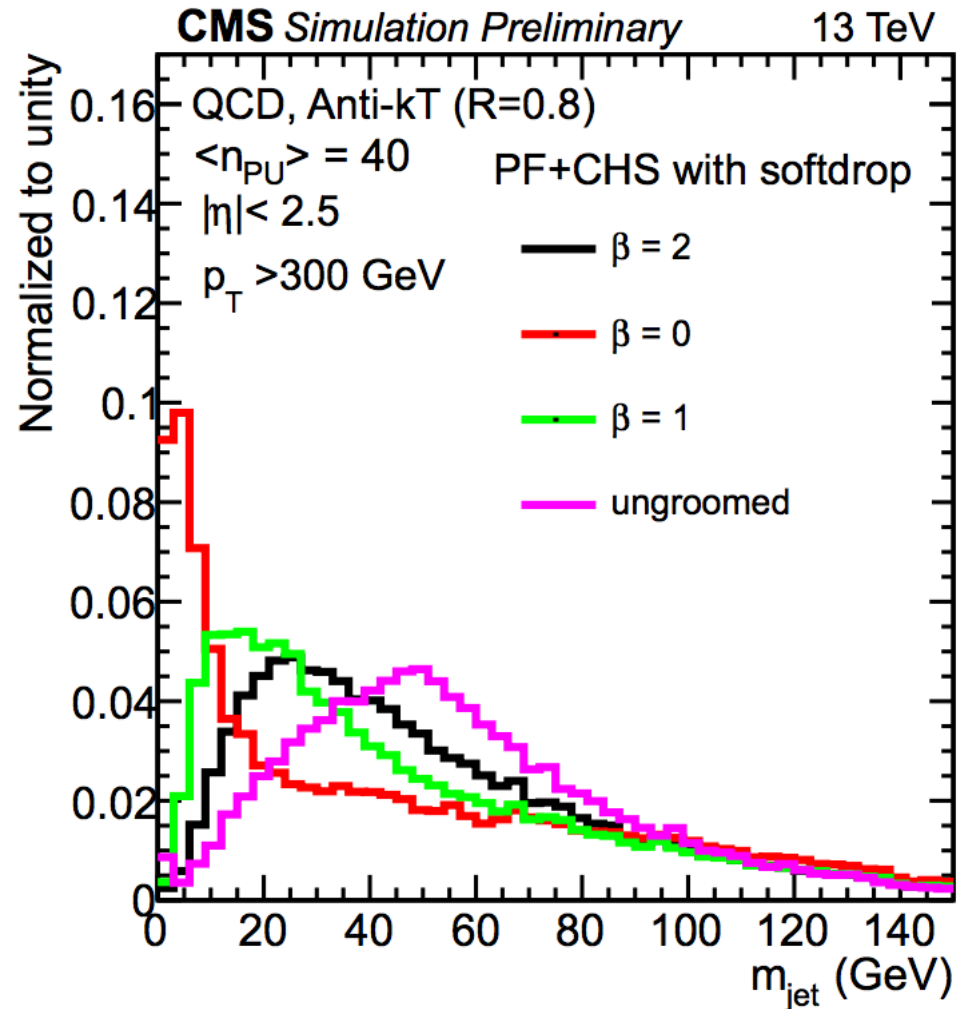
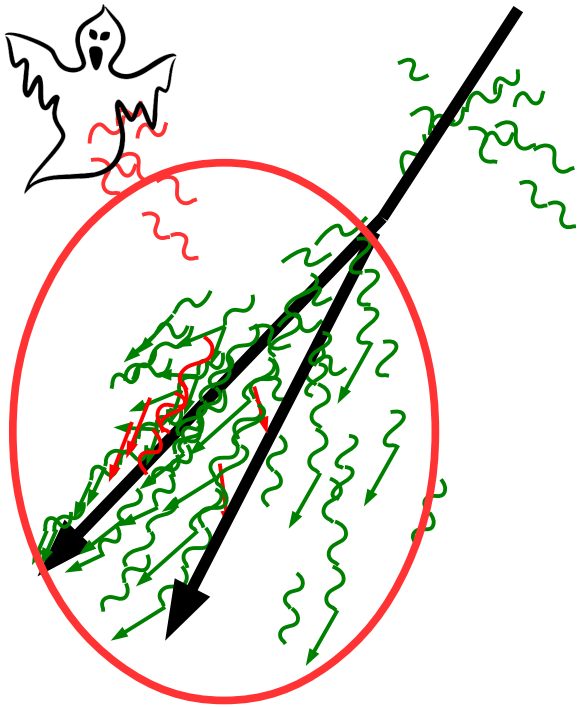


In the case about the individual process of the jet

This will be focus of the rest of this talk

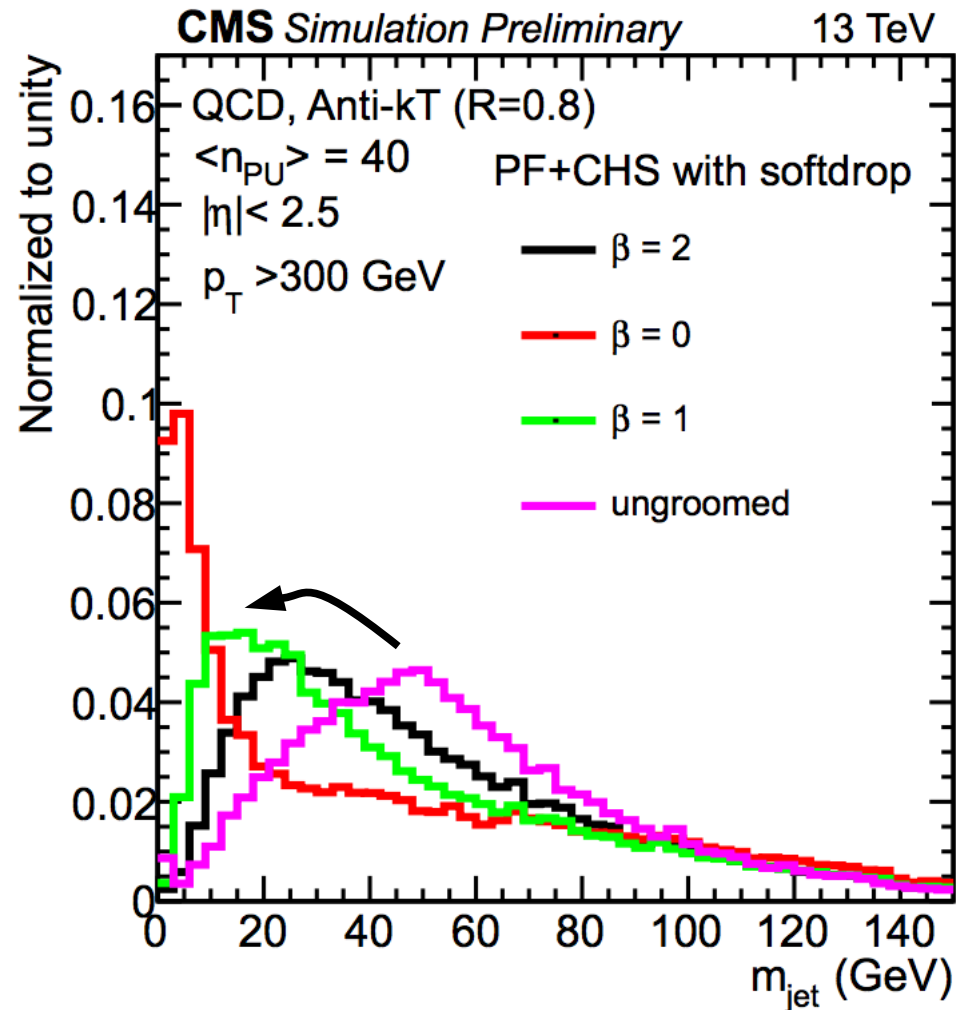
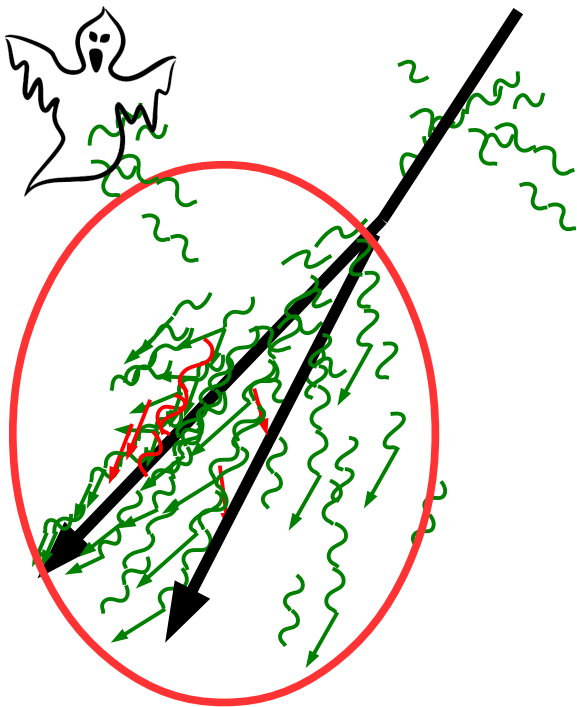
Jet Mass Measurement

- Improving the measurement



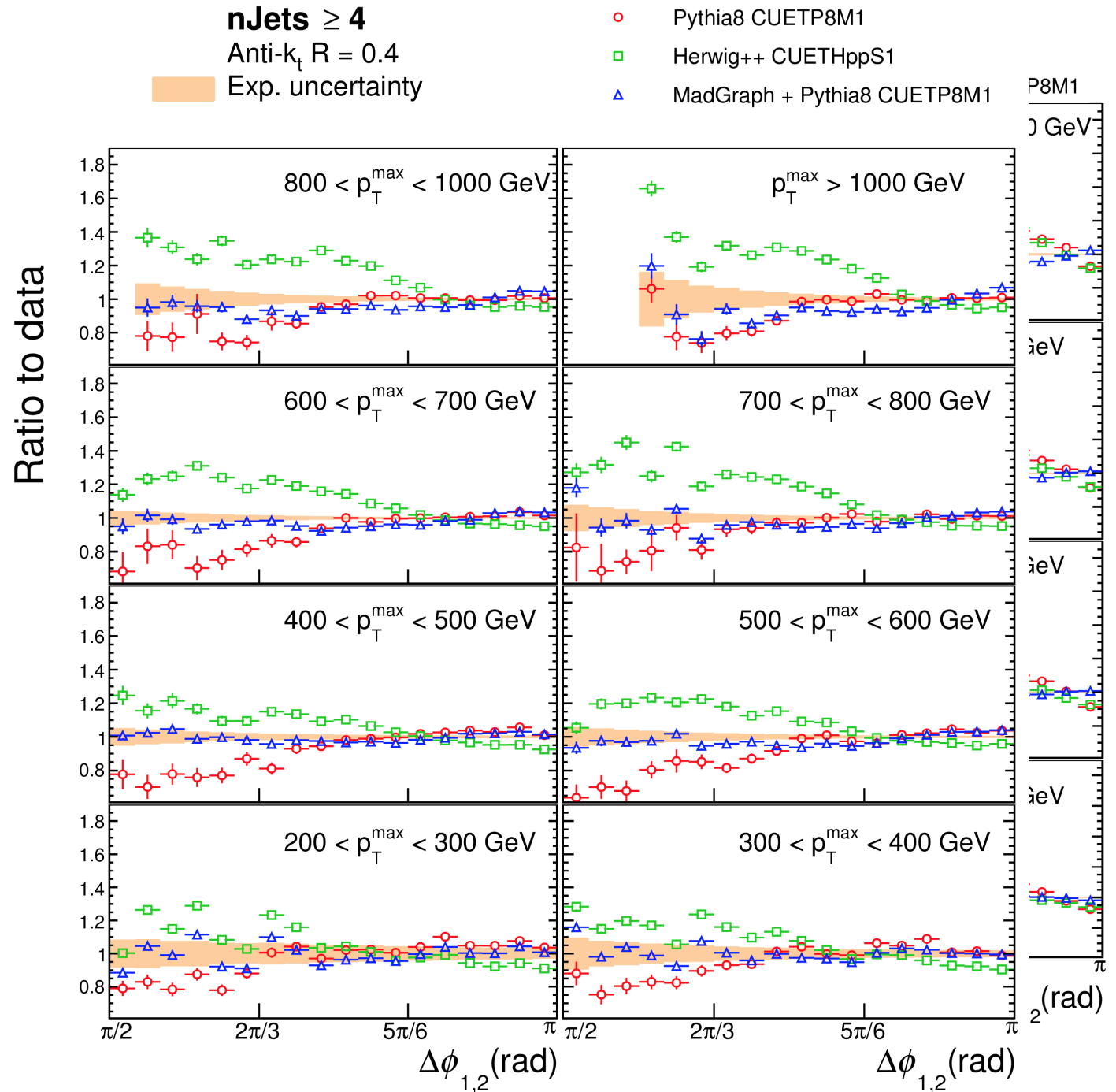
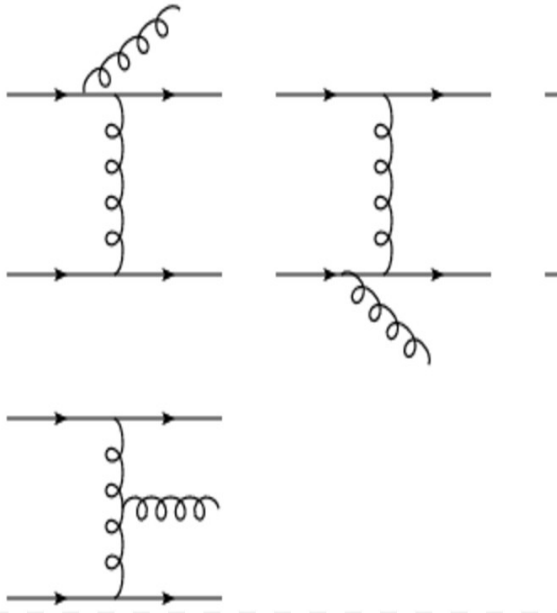
Jet Mass Measurement

- Improving the measurement

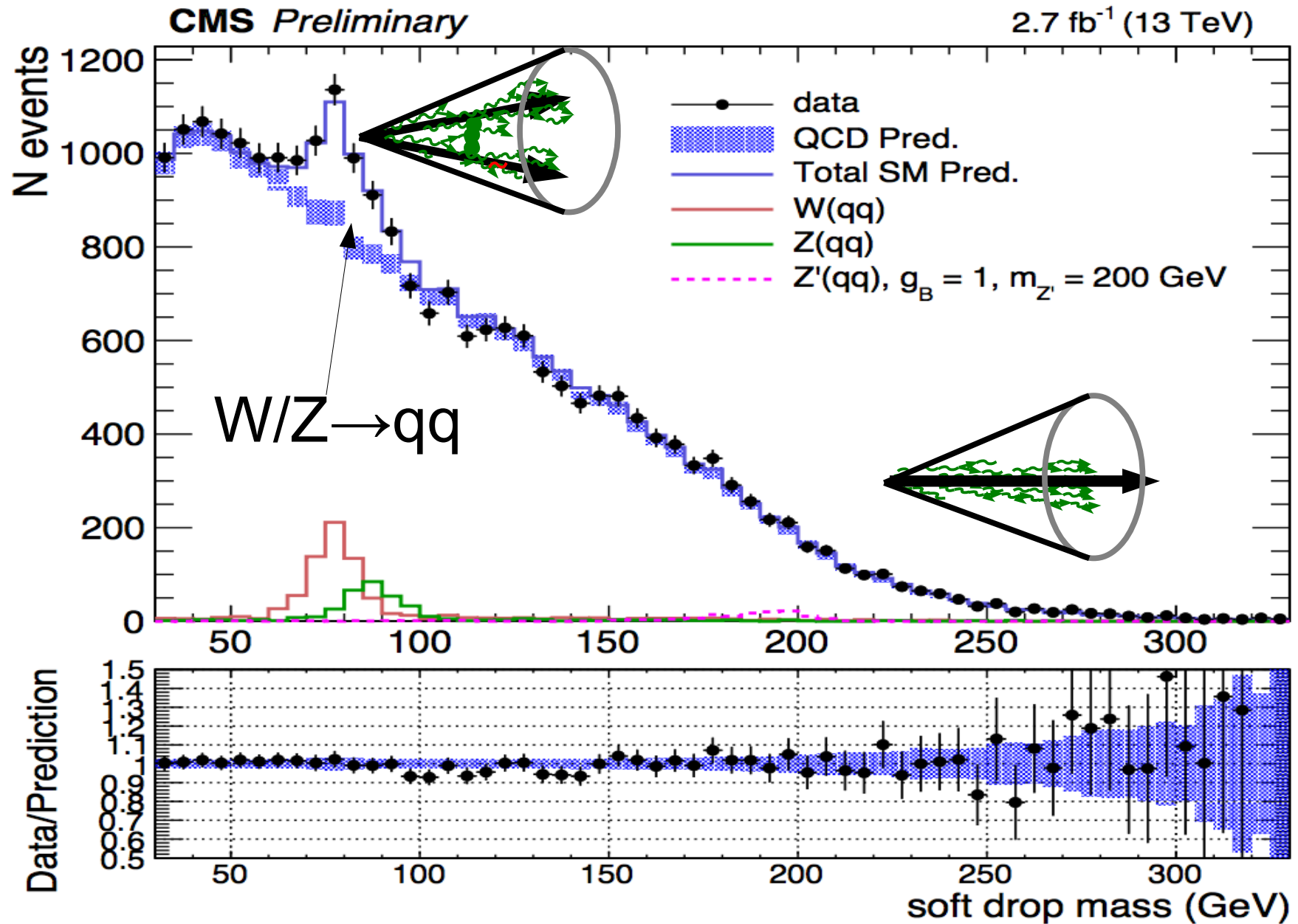


In slightly less old
Cared about 3]

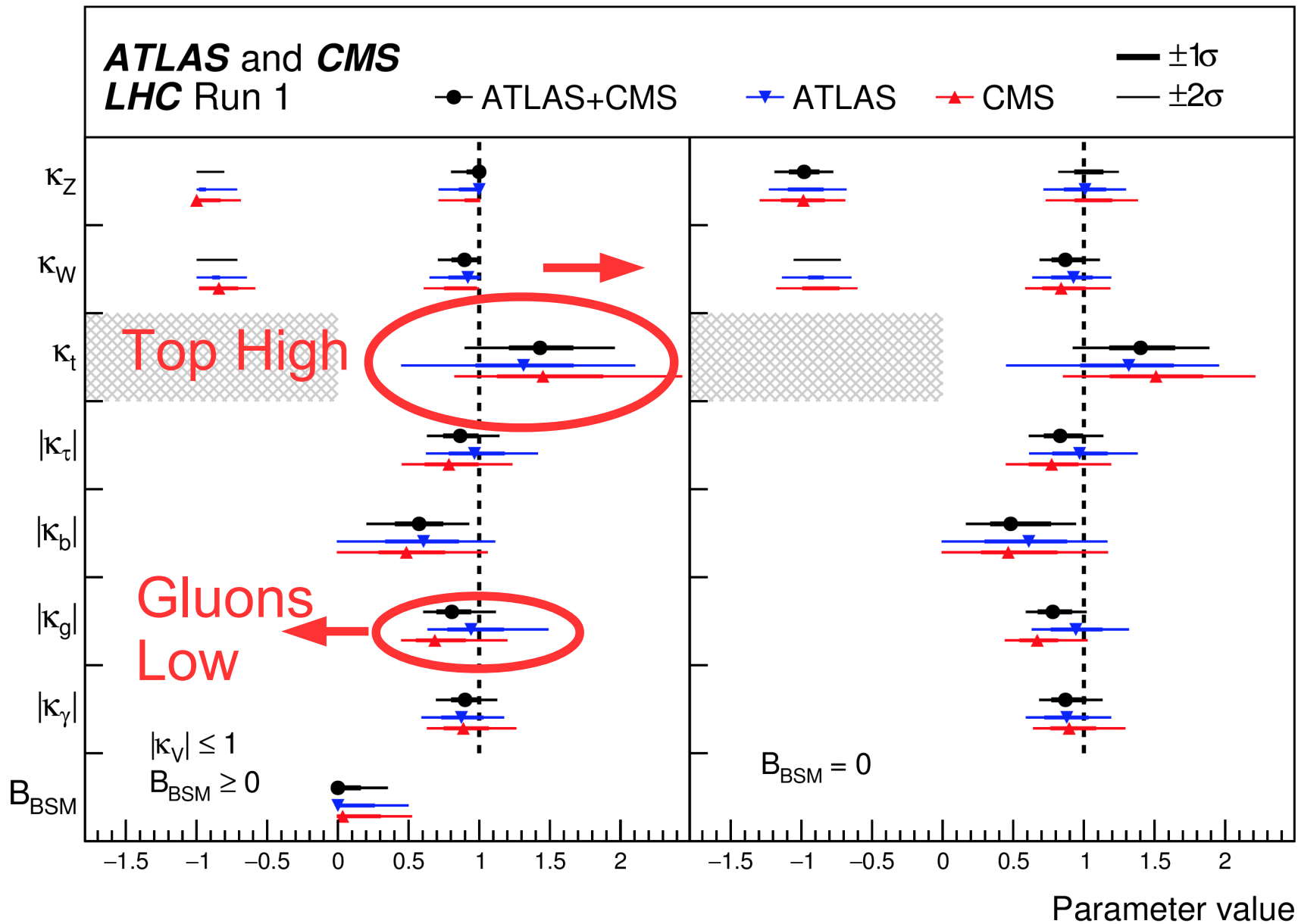
The concern here
Split out parton
study additional
properties



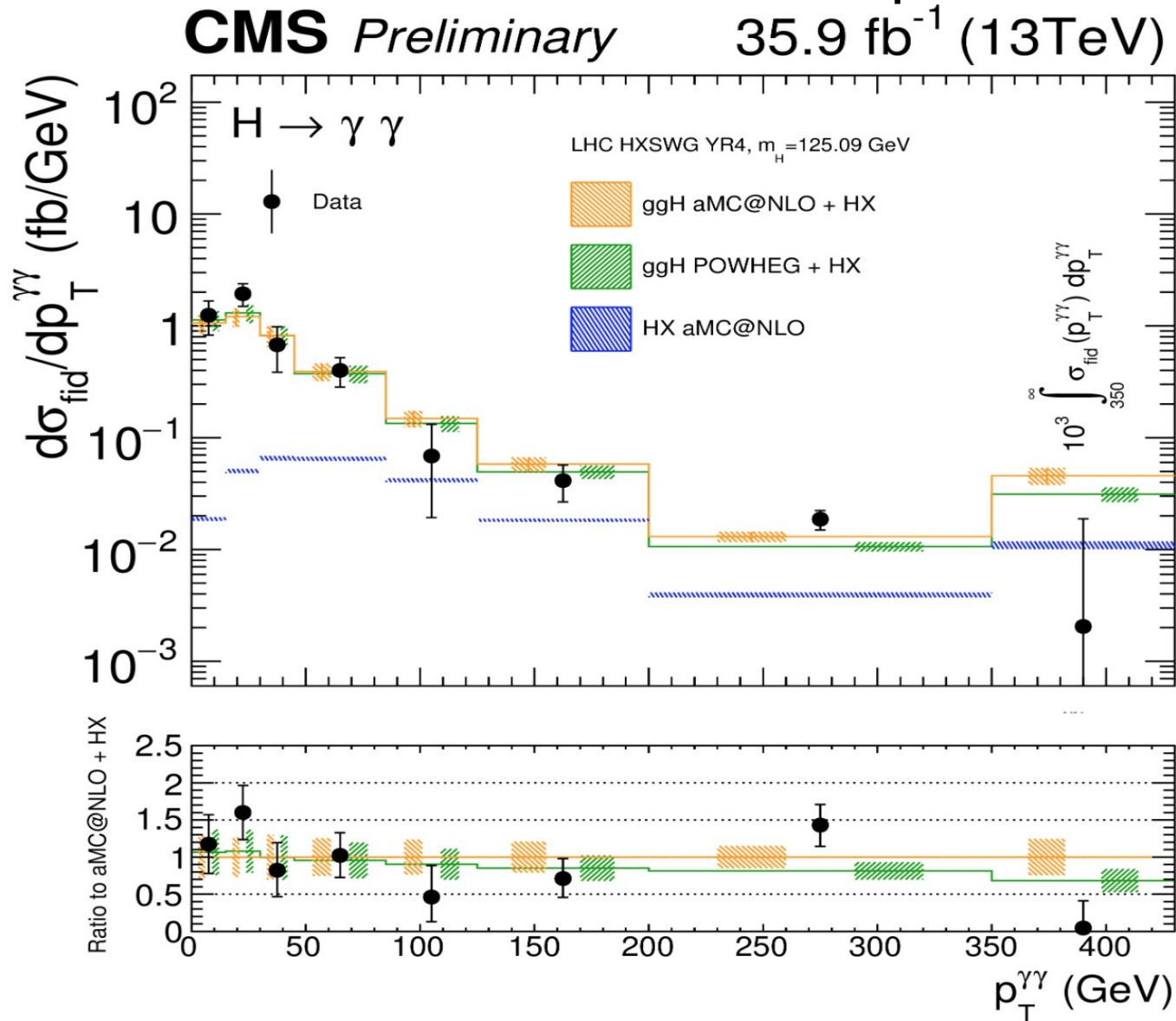
Jet Mass distribution



Hints?



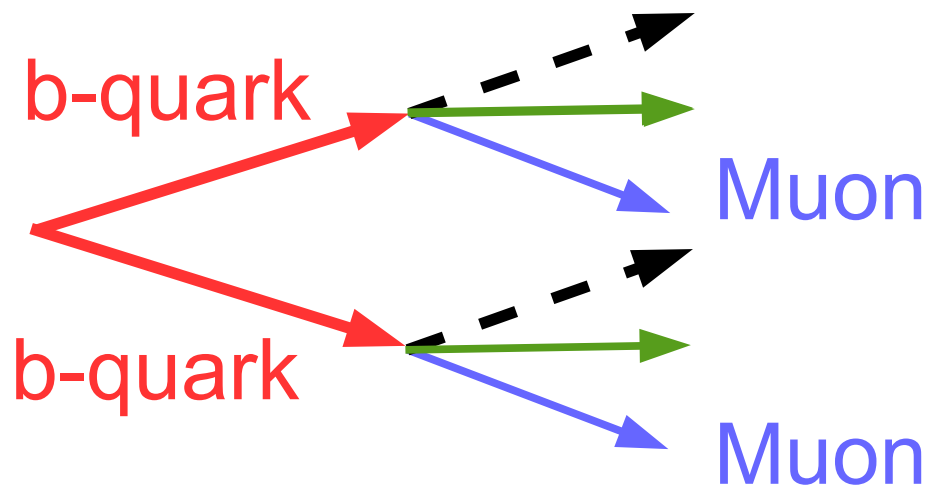
Current Best Higgs p_T measurement



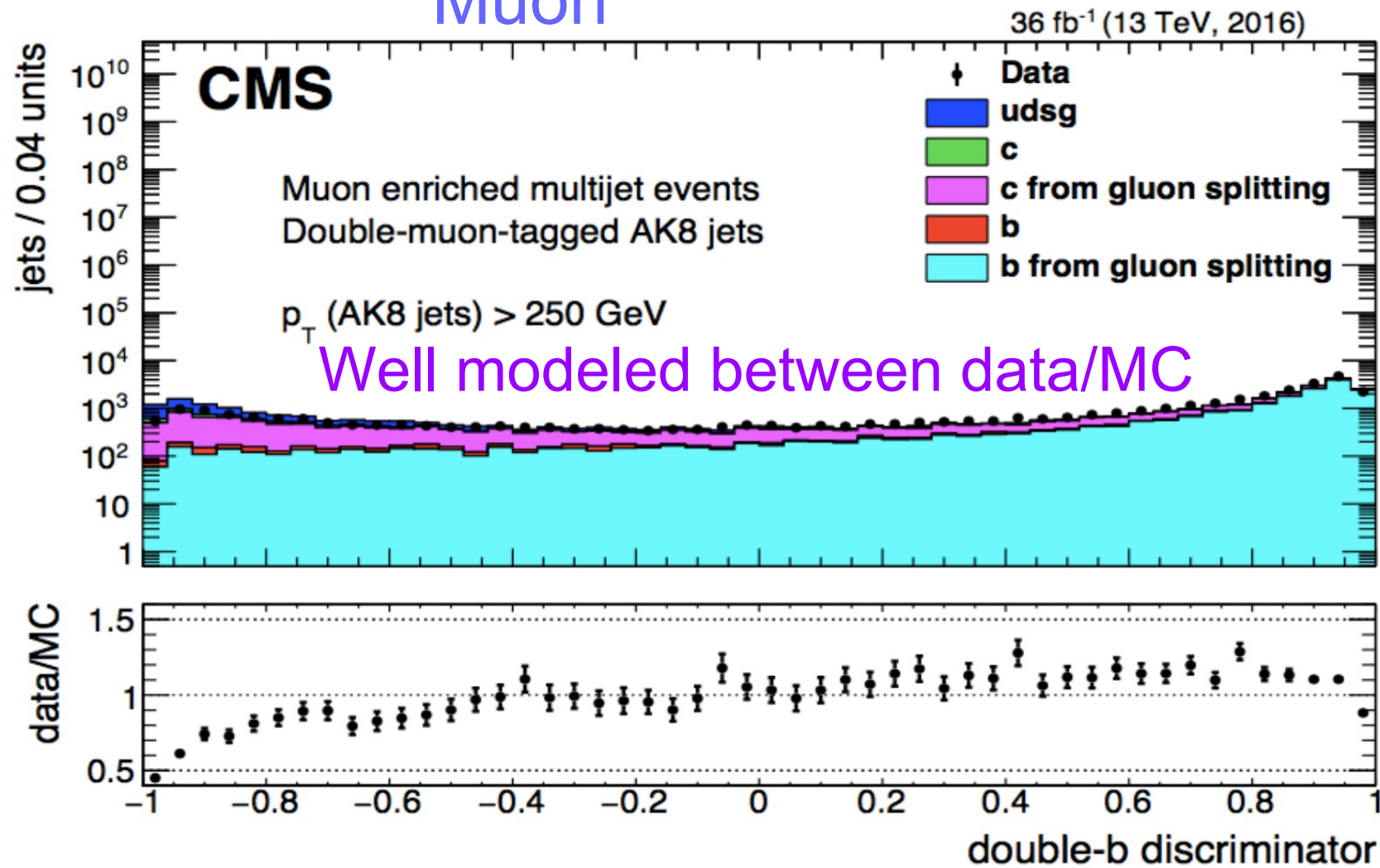
Analysis
starts
here

No other way to look for events in this region

What about data/MC

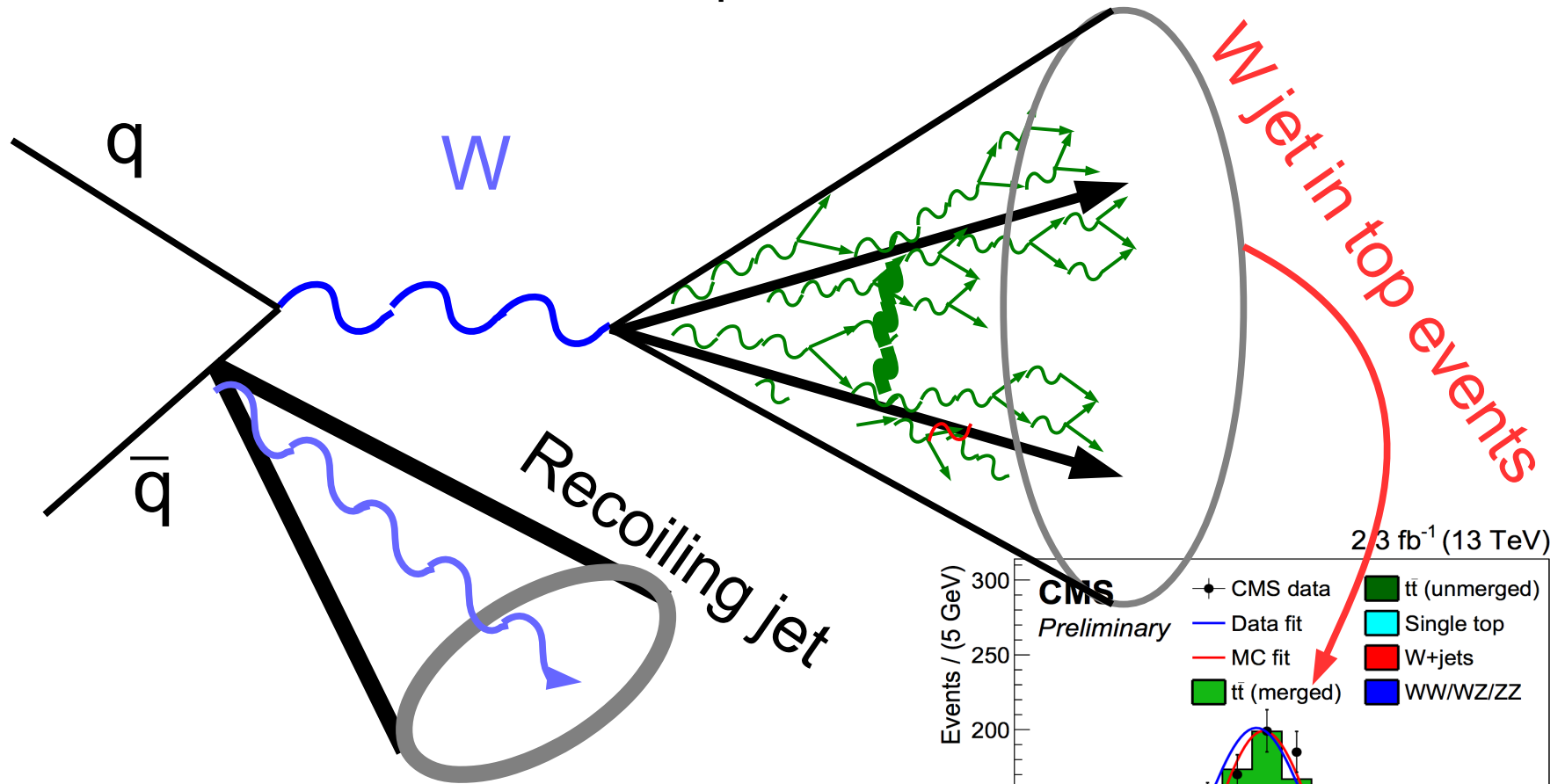


Tag two muons in a jet
Use this to infer signal-like
2 b -quarks in a jet

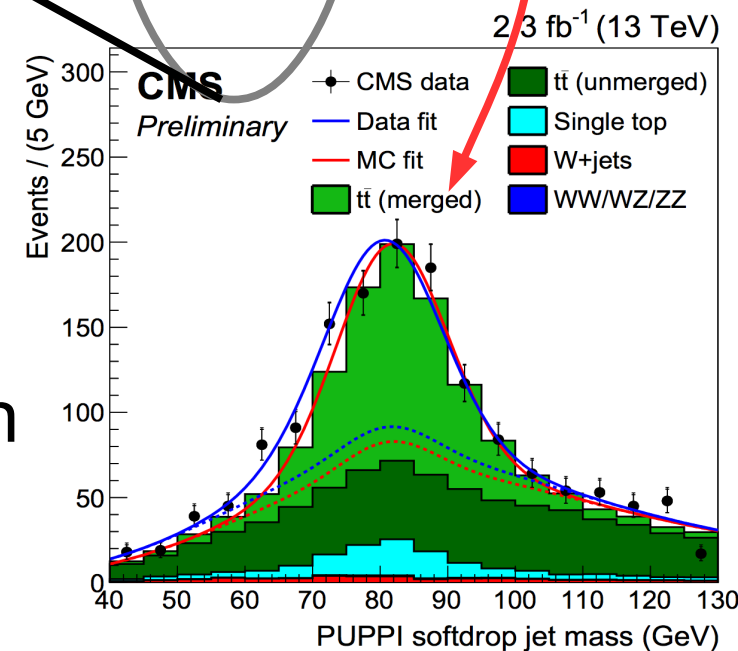


Whats the first goal?

Require W to be a high p_T

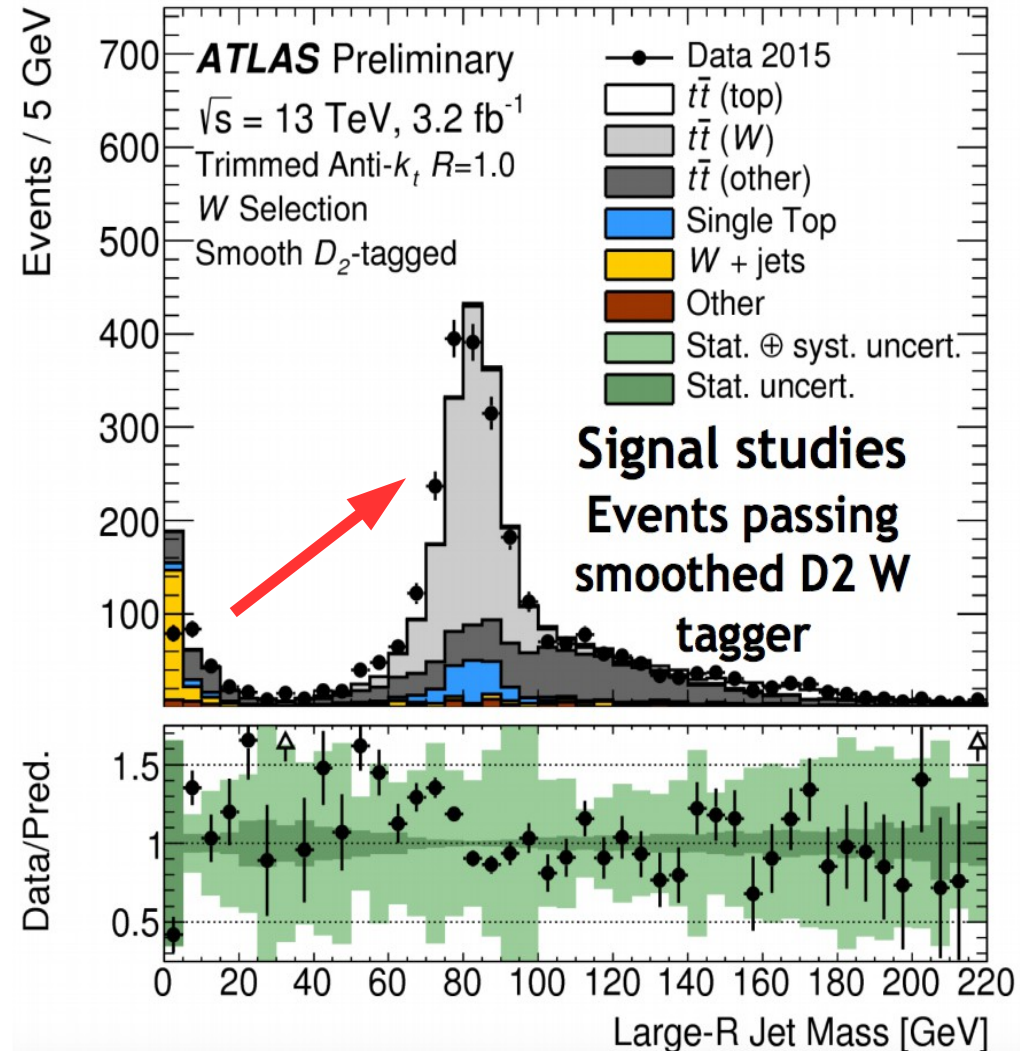
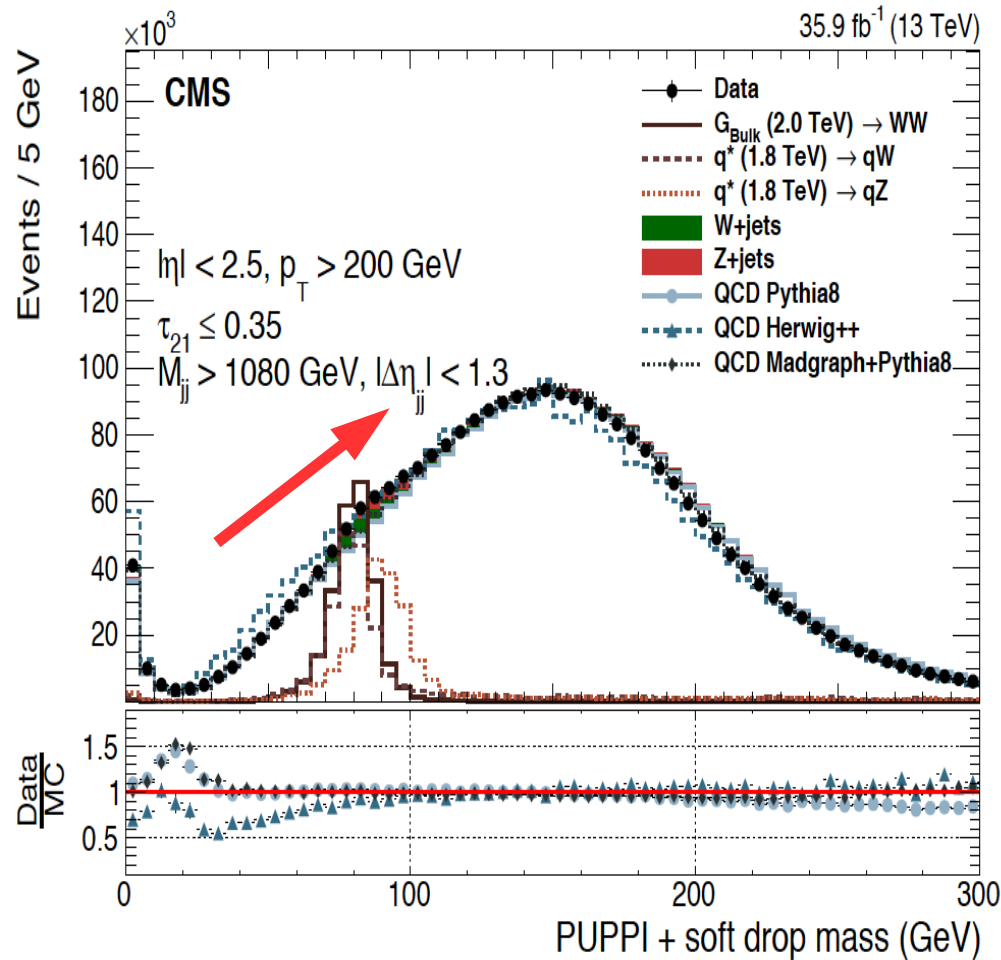


Have not observed a single W jet boson event from SM produce
(Have only seen W jets in top events)



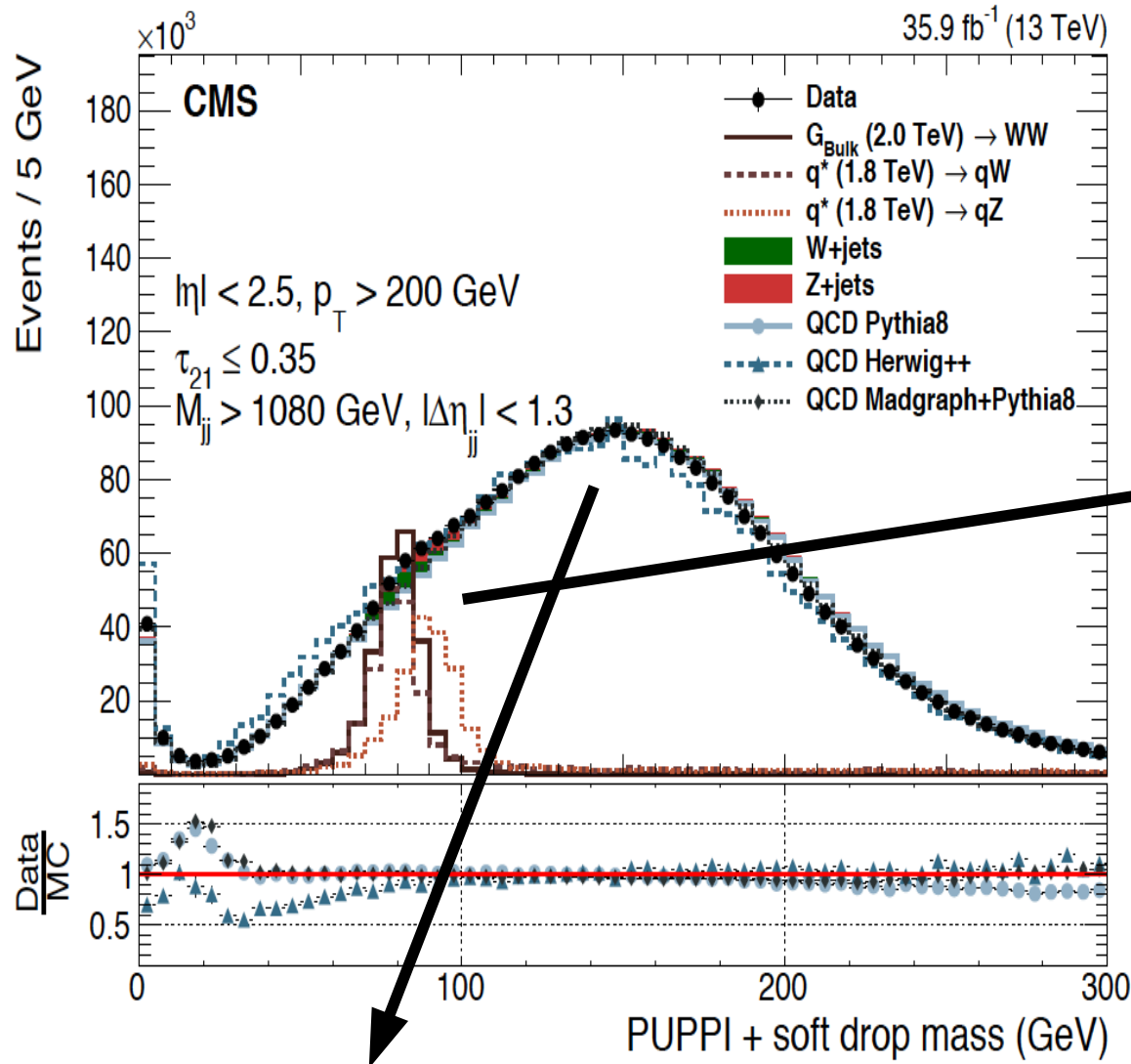
.....at the same time on CMS

- An observation was being made



Background with current cuts force an upward slope

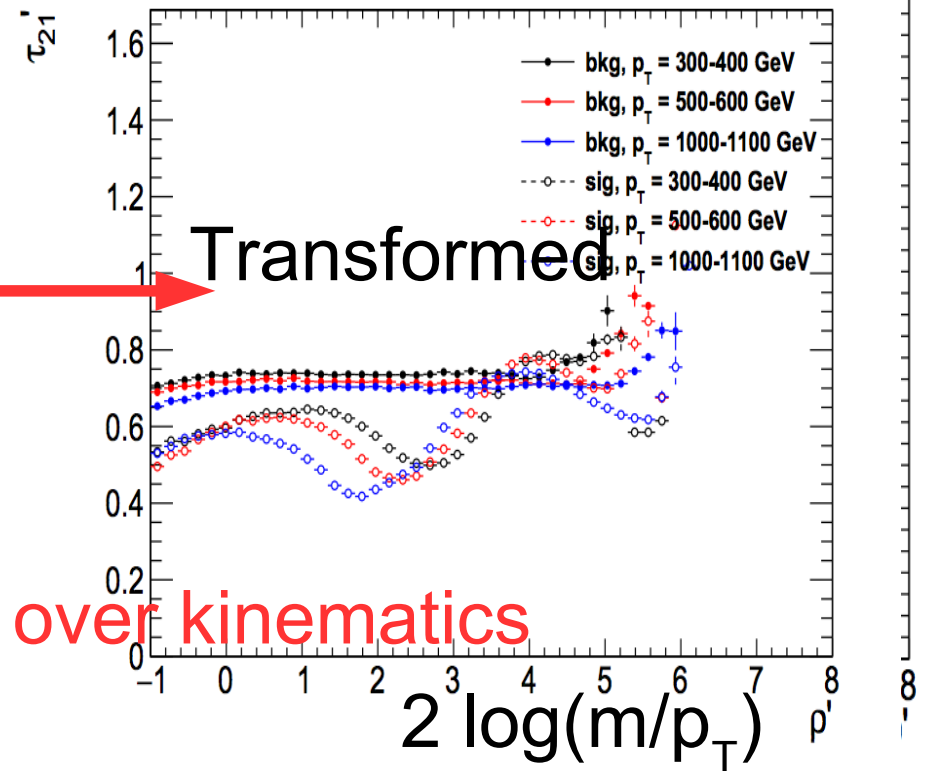
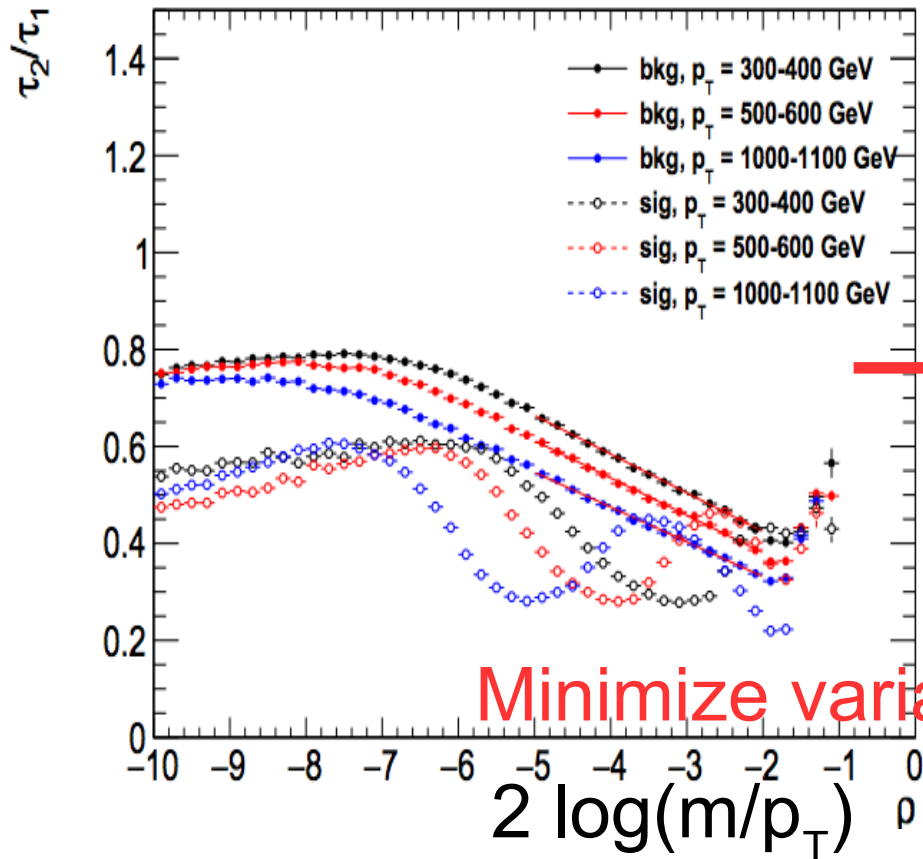
Finding the W/Z on a slope



How does the upward slope impact the search for a resonance?

Background is peaking after the bump
 helps to improve reduce background at W

Designing a new tagger?



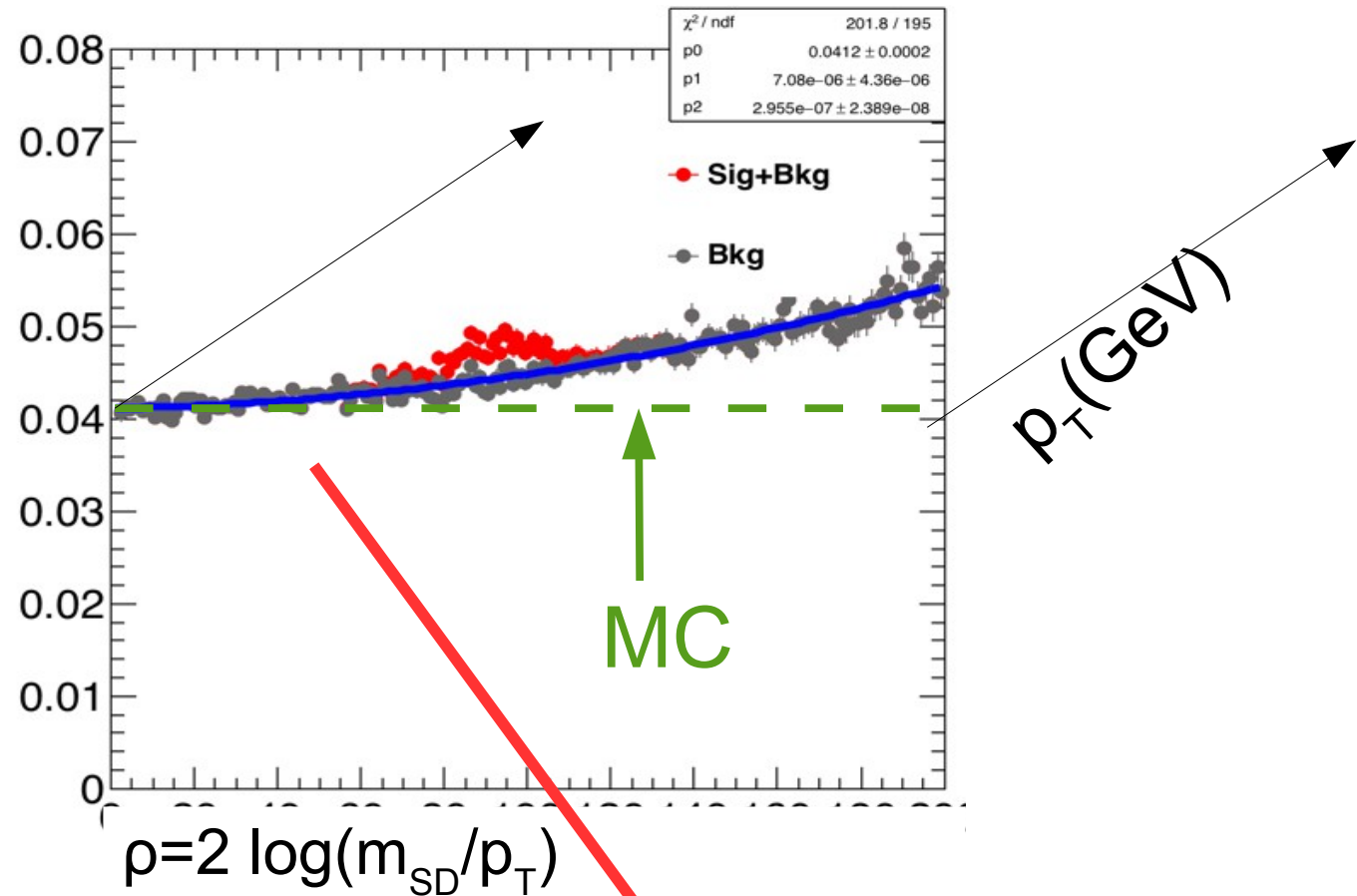
$$\tau_{21}' = \tau_2/\tau_1 - M \times \rho'$$

$$\rho' = \rho + \log \frac{p_T}{\mu} = \log \left(\frac{m^2}{p_T \mu} \right)$$

To minimize sculpting we minimize the variation over mass and p_T

In data

- We can tune the efficiency to be flat in MC

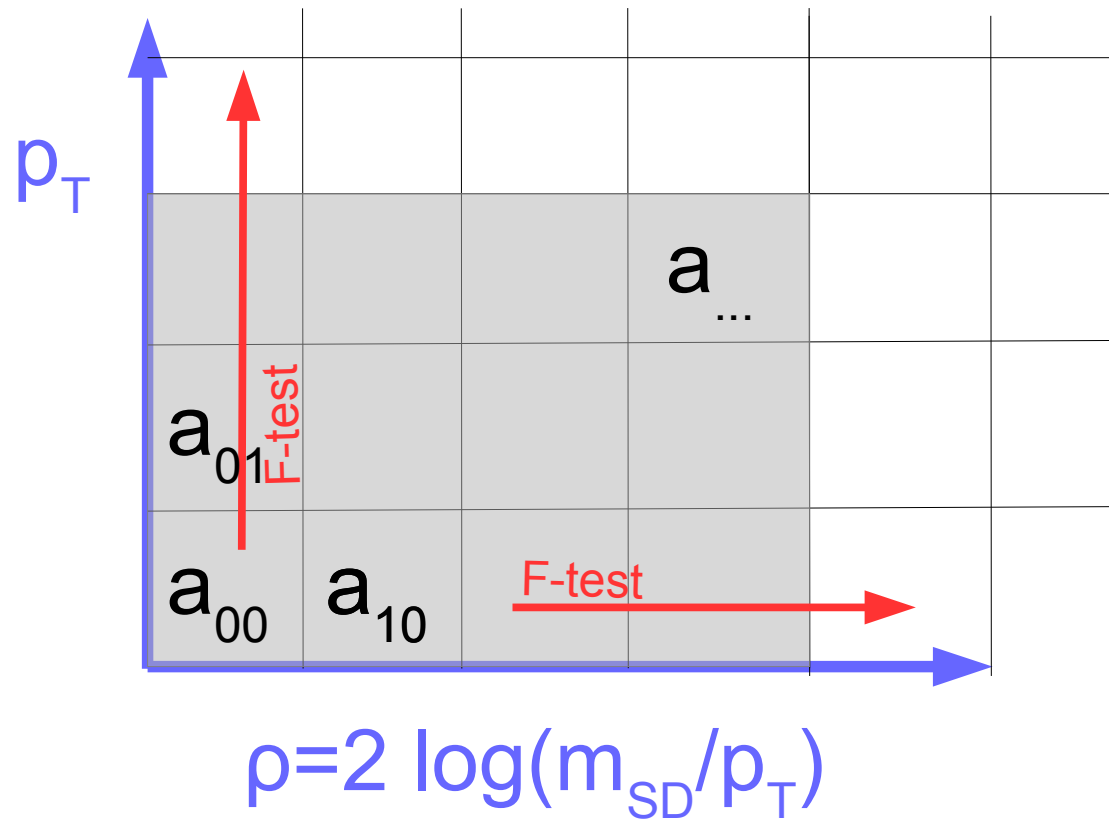


Utilize a polynomial in p and p_T

$$N_{\text{pass}}^{\text{QCD}}(m_{SDi}, p_{Tj}) = \epsilon^{\text{QCD}} \cdot \left(\sum_{k,\ell} a_{k\ell} \rho_{ij}^k p_{Tj}^\ell \right) \cdot N_{\text{fail}}^{\text{QCD}}(m_{SDi}, p_{Tj})$$

In data

- Interactively expand in ρ & p_T order with an f-test



For W peak
Using
4th order in ρ
And
3rd order in p_T

Utilize a polynomial in ρ and p_T

$$N_{\text{pass}}^{\text{QCD}}(m_{SDi}, p_{Tj}) = \epsilon^{\text{QCD}} \cdot \left(\sum_{k,\ell} a_{k\ell} \rho_{ij}^k p_{Tj}^\ell \right) \cdot N_{\text{fail}}^{\text{QCD}}(m_{SDi}, p_{Tj})$$

Ingredients to see W peak: Recp

- Select a jet ($p_T > 500$)
- Require the jet to pass a 2 prong tag
 - Tune the tagger to be flat across the mass
- Select tagger & take both pass and failing selection
- Simultaneously fit for pass and failing
- Parametrize the difference between data and MC
- Use a 4th , 3rd order polynomial in p, p_T to model data

Now are we ready to fit the W peak?

The End Game

Meanwhile...

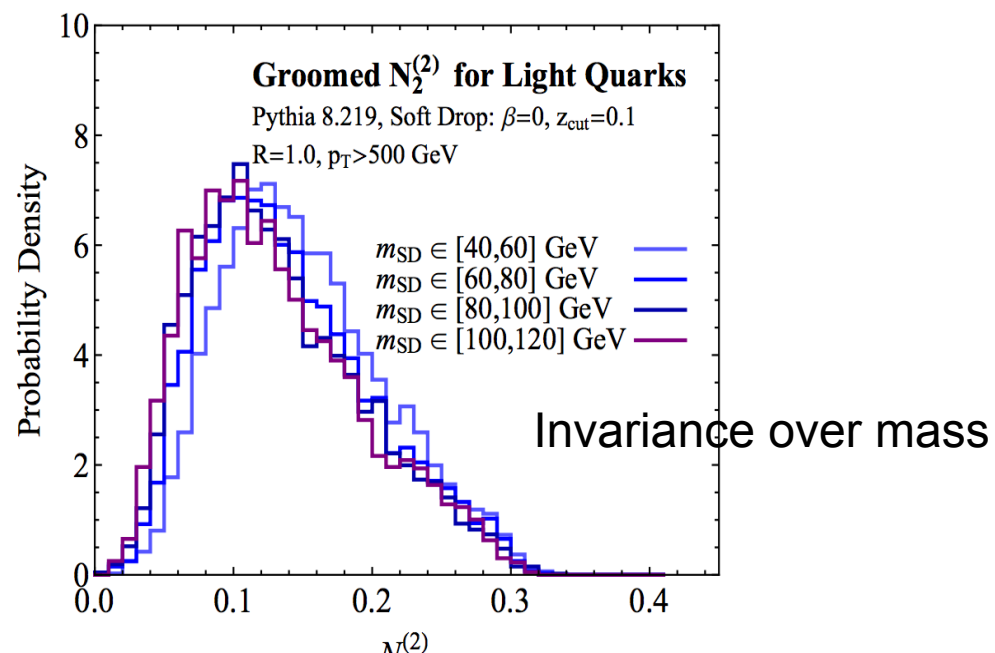
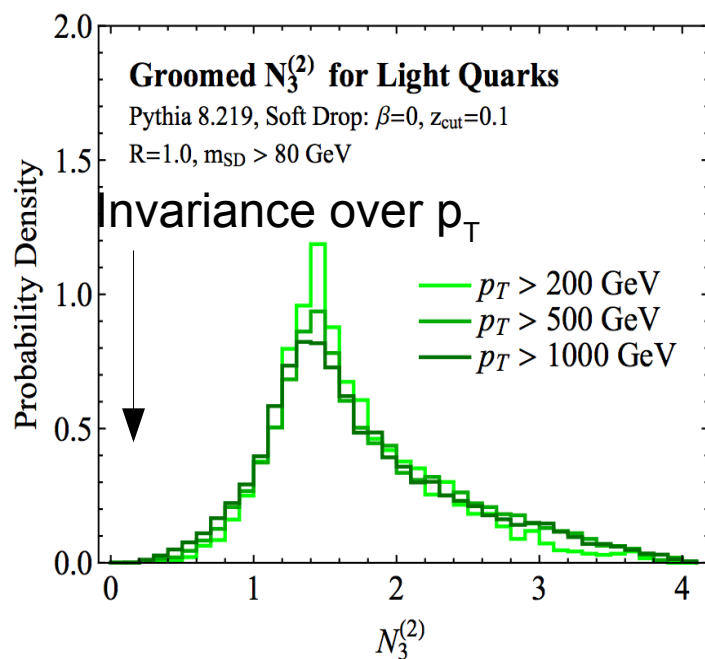
- Theorists decided to build in the scale invariance

New Angles on Energy Correlation Functions

Ian Mout, Lina Necib, Jesse Thaler

(Submitted on 23 Sep 2016)

- Into a new set of substructure observables
- Guiding principles are to exploit invariances in QCD



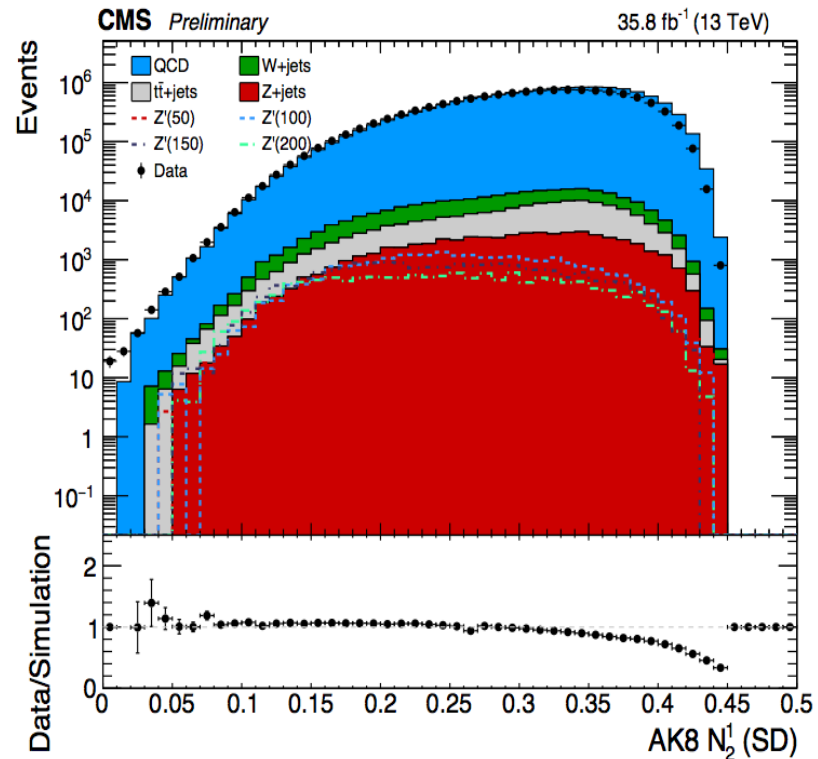
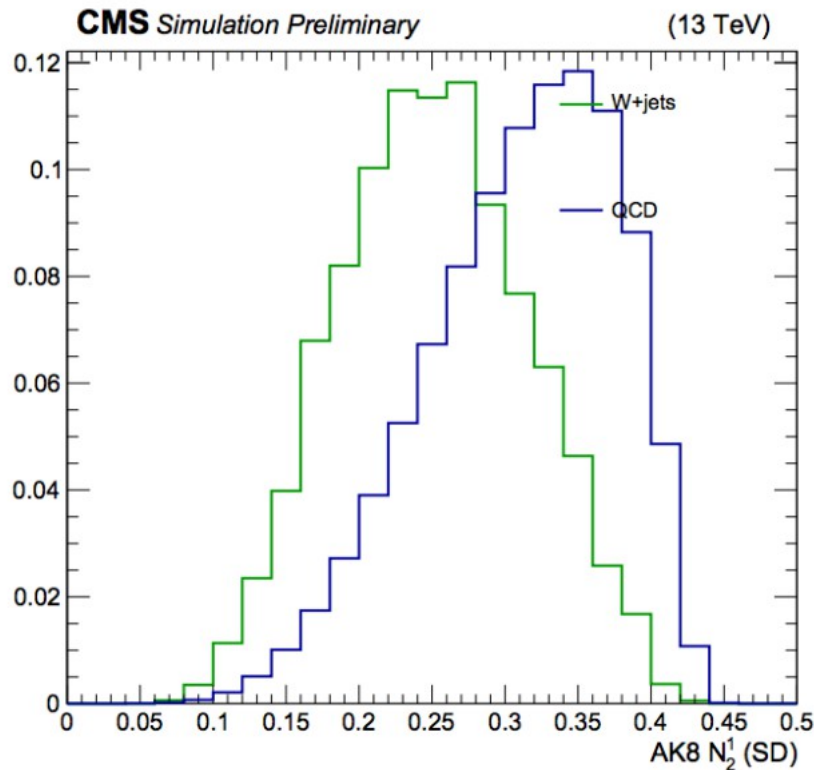
New Substructure Observables

$$N_2(\beta) = \frac{2e_3^\beta}{(e_2^\beta)^2}$$

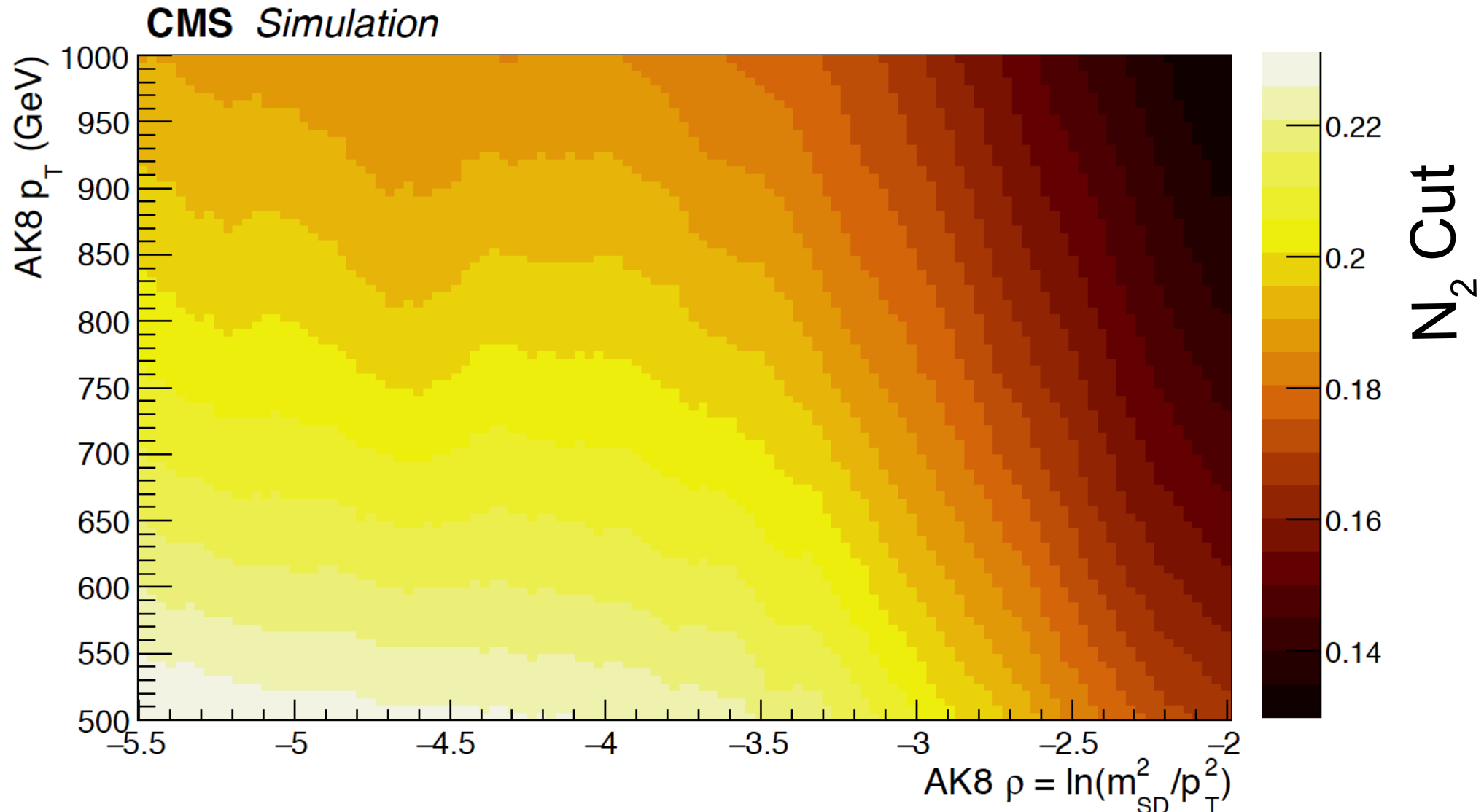
$$1e_2^\beta = e_2^\beta = \sum_{1 \leq i < j \leq n_l} z_i z_j \Delta R_{ij}^\beta \quad z_i \equiv \frac{p_{Ti}}{\sum_{j \in \text{jet}} p_{Tj}}$$

$$2e_3^\beta = \sum_{1 \leq i < j < k \leq n_l} z_i z_j z_k \min\{\Delta R_{ij}^\beta \Delta R_{ik}^\beta, \Delta R_{ij}^\beta \Delta R_{jk}^\beta, \Delta R_{ik}^\beta \Delta R_{jk}^\beta\}$$

Using AK8 PUPPI jets



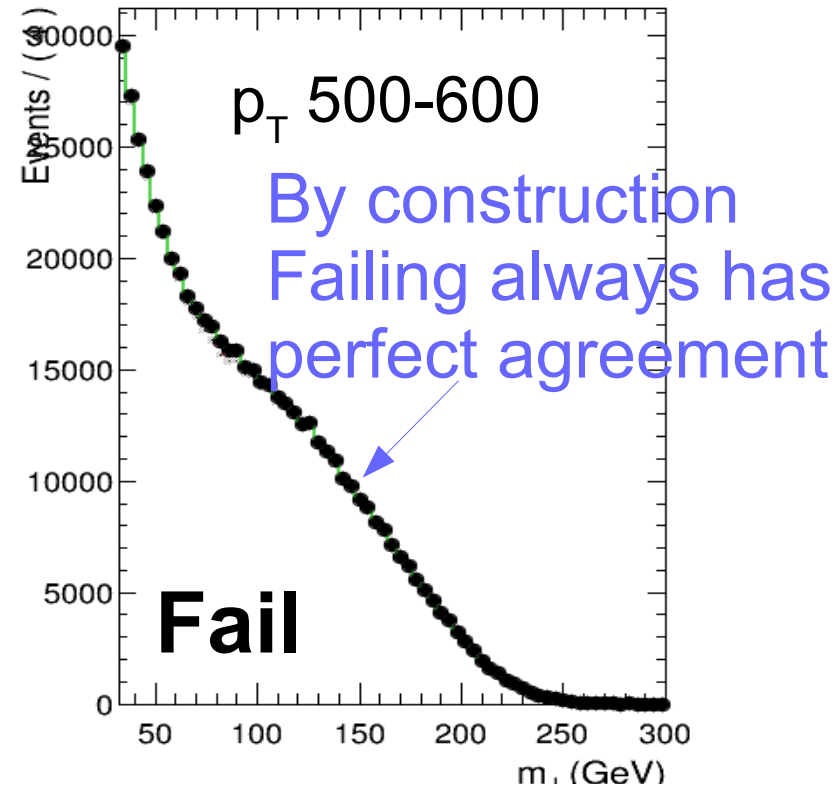
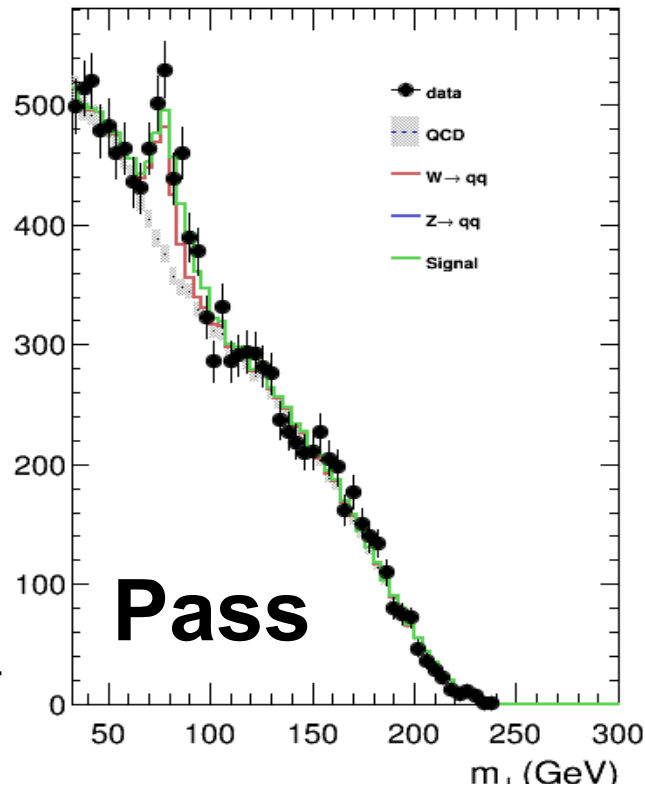
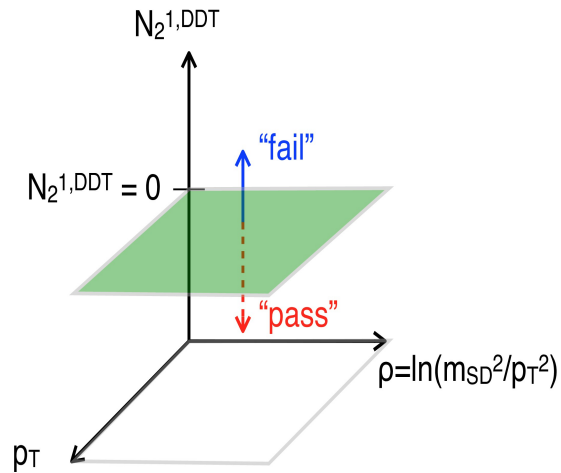
Full Decorrelation scheme



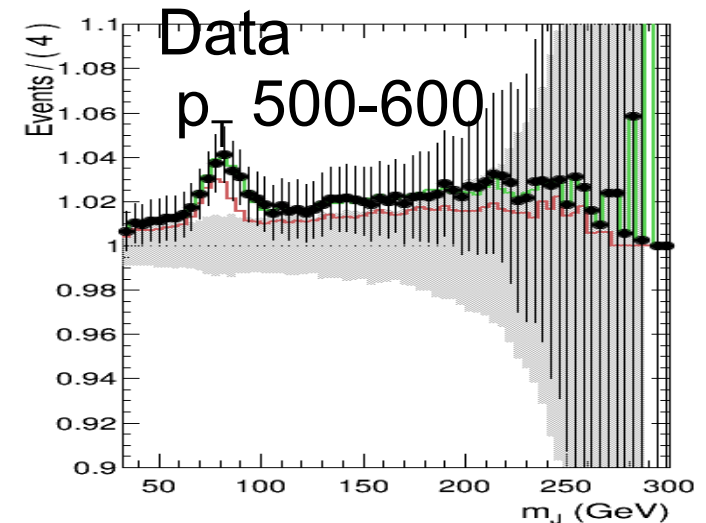
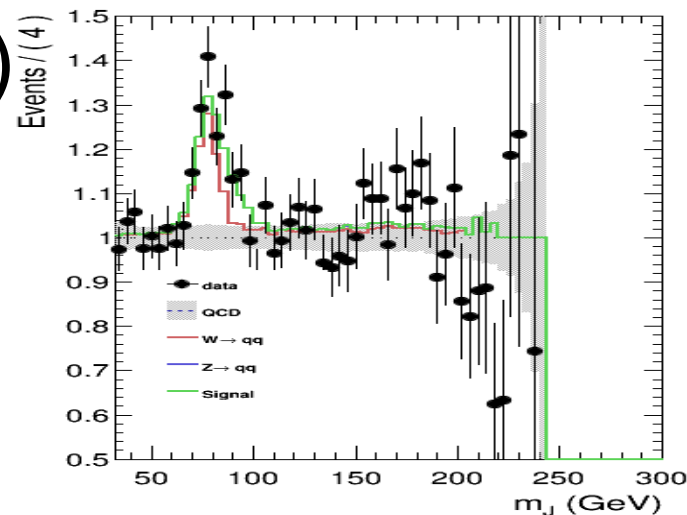
Use the k-Nearest Neighbor approach to determine N_2 cut

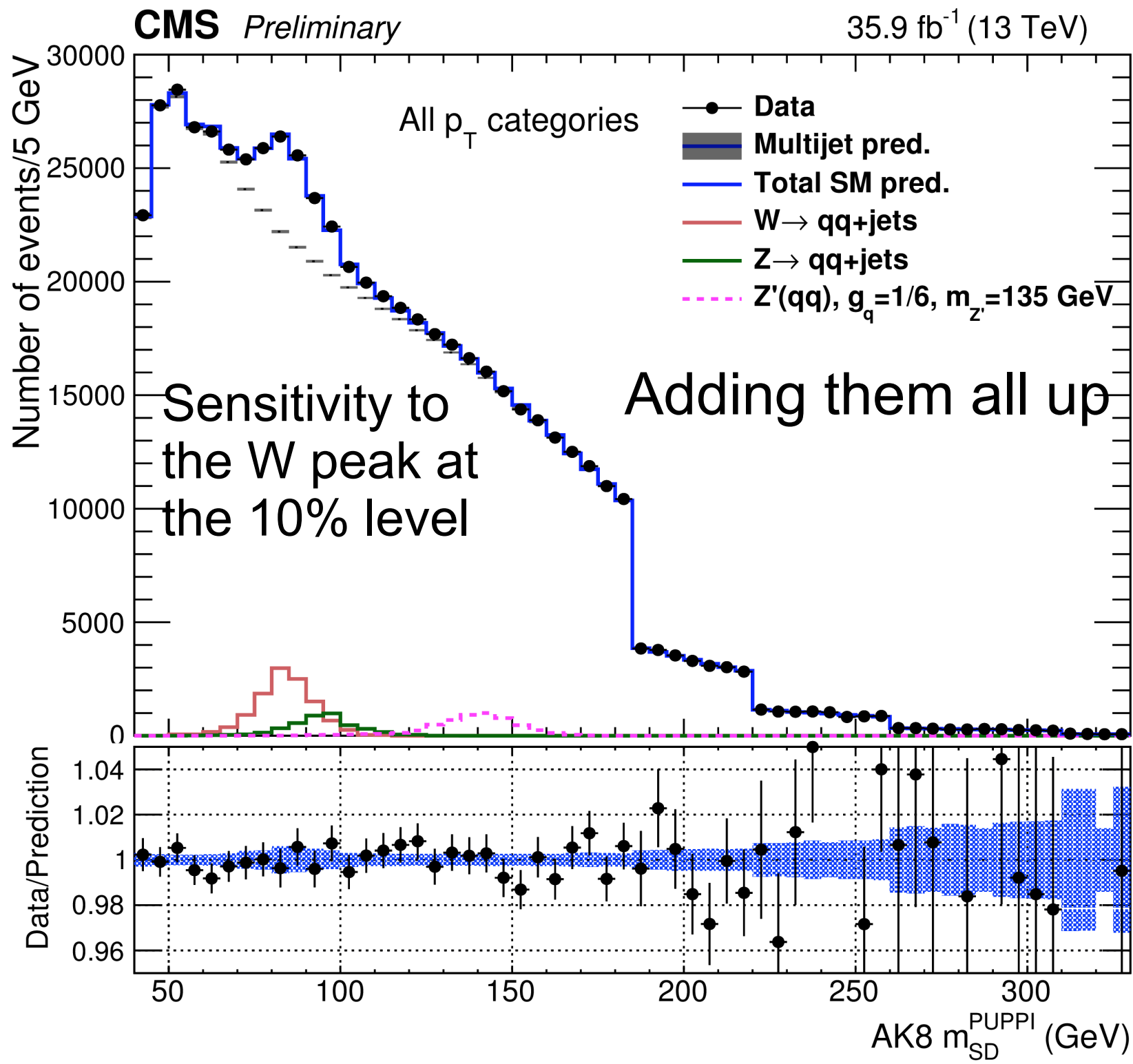
All Pieces are in Place

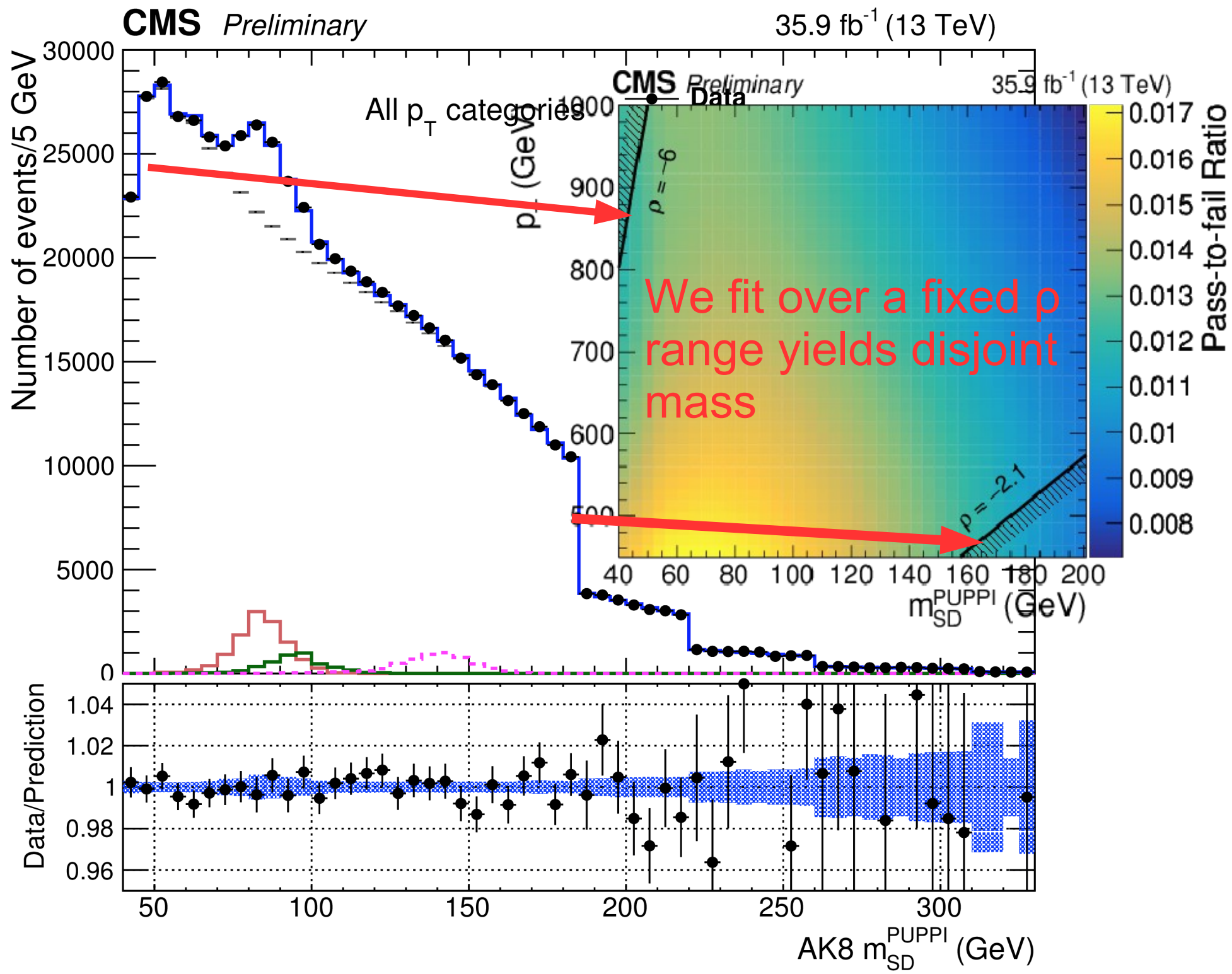
- The final fit



Use 5 bins in p_T
500-1000 GeV
(100 GeV bins)

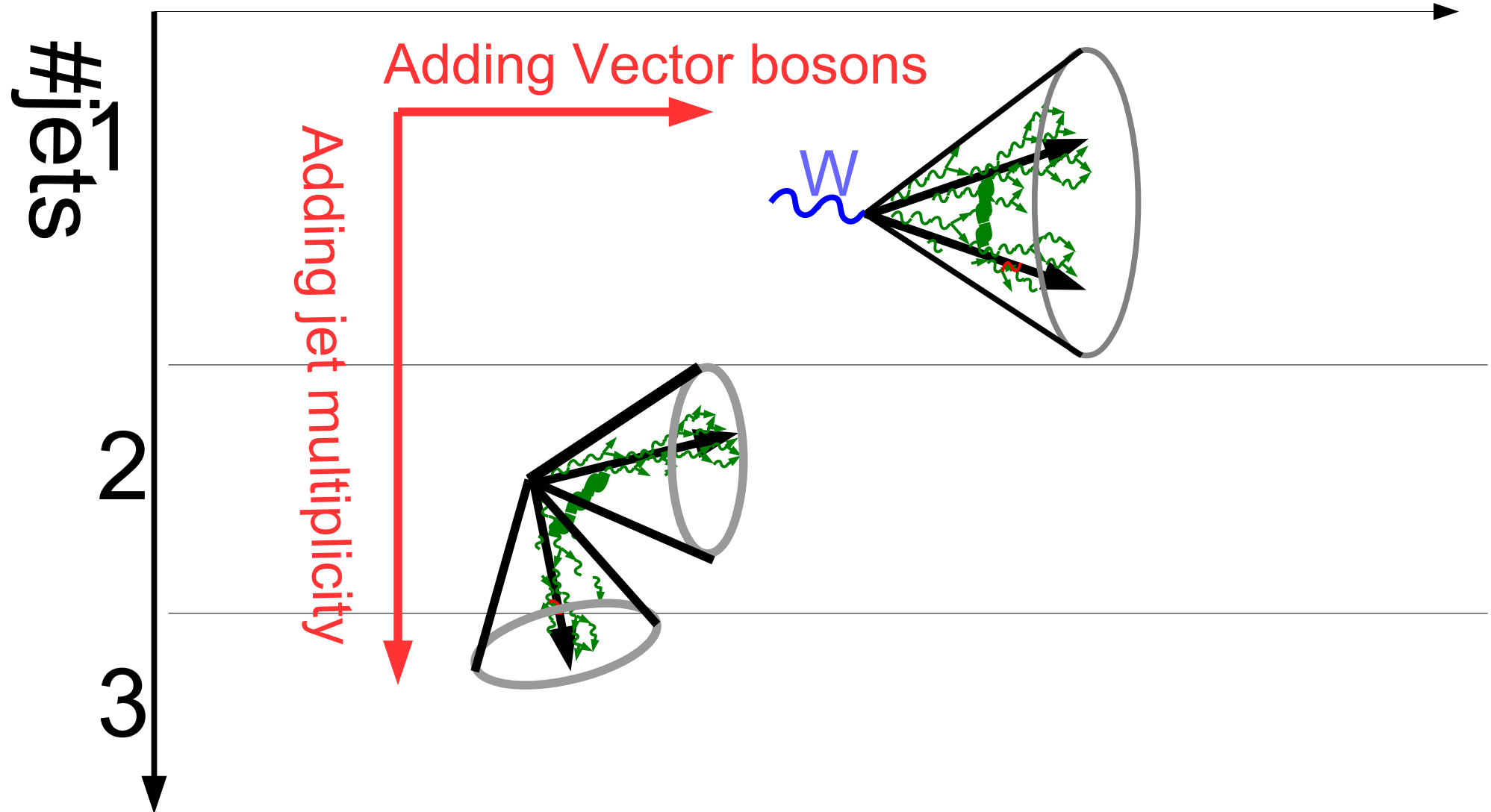






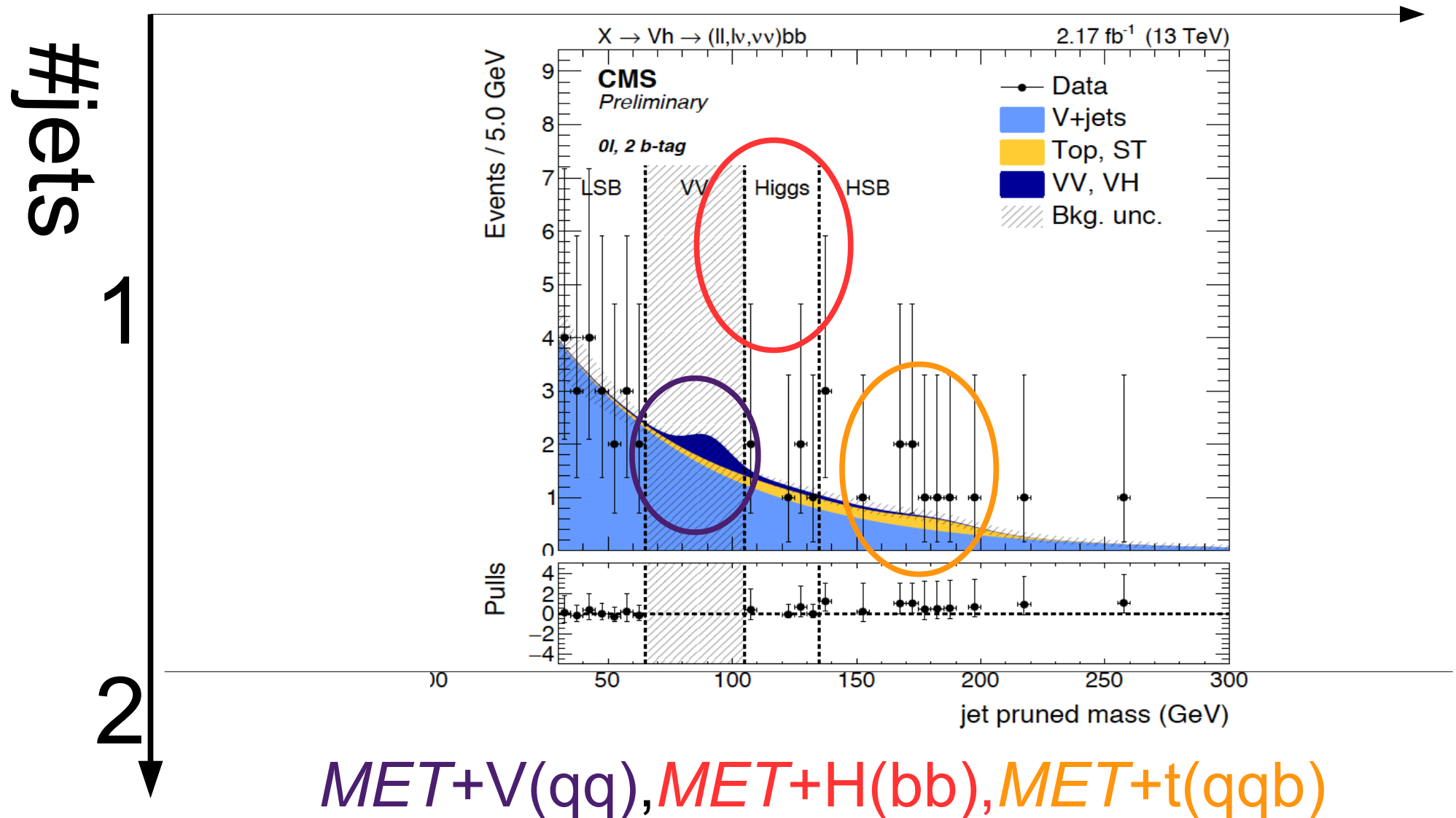
Monojet Extension Plane

Jet Mass



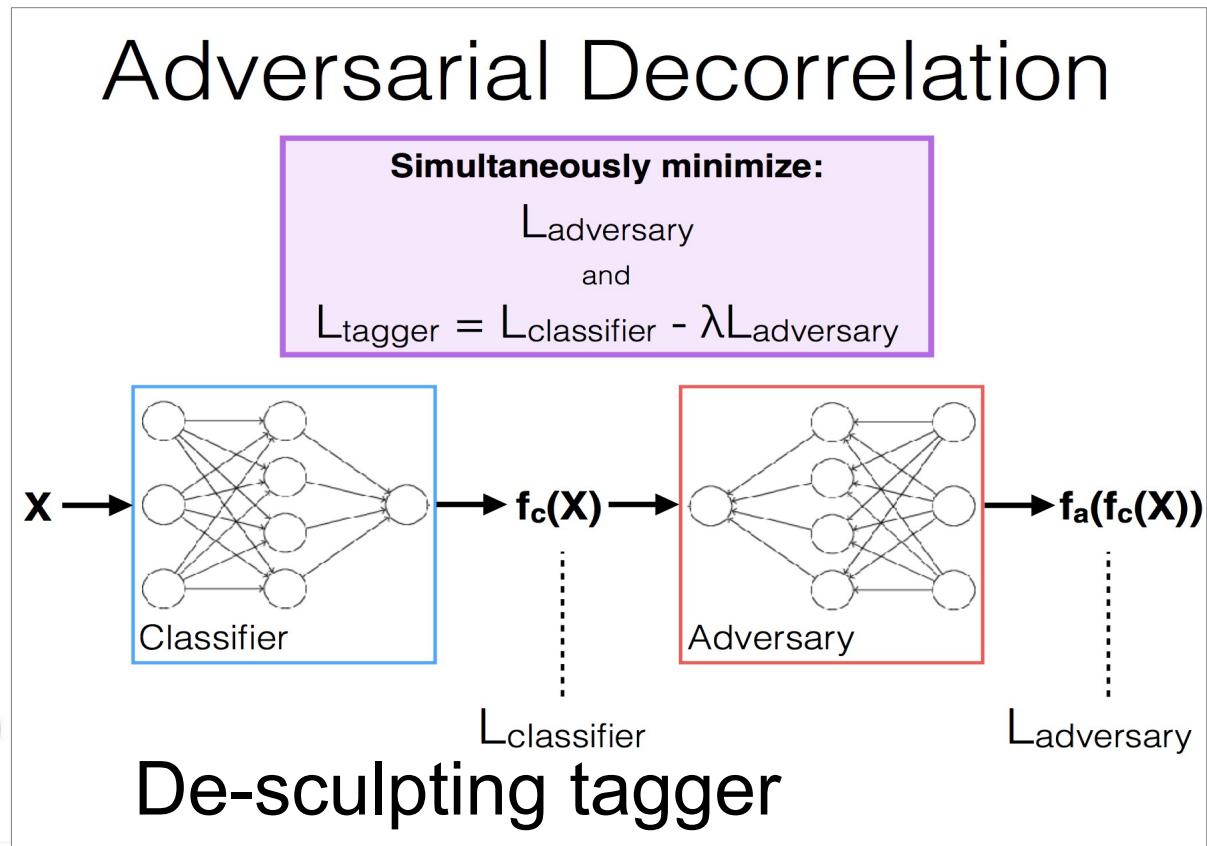
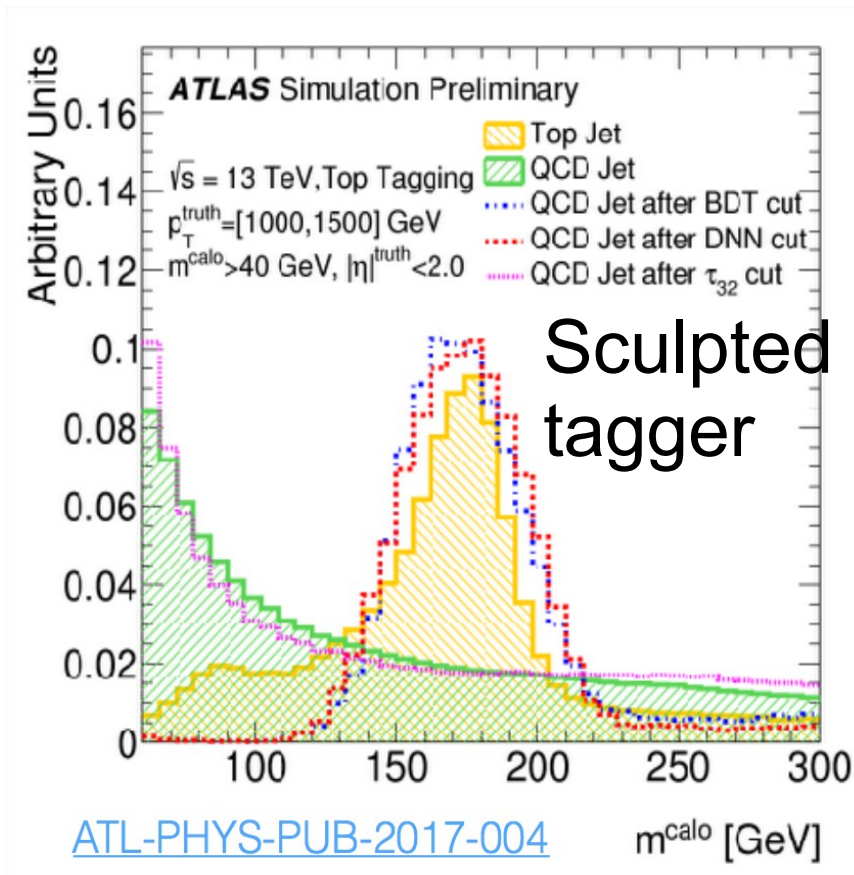
Monojet Extension #1 ($V \rightarrow qq$)

Jet Mass



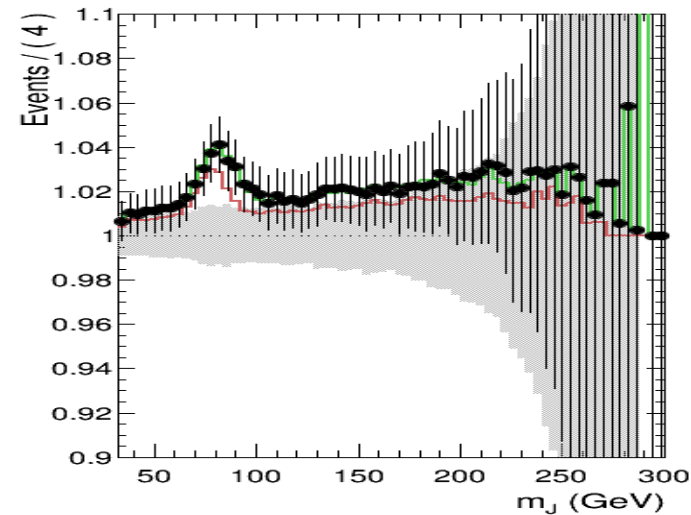
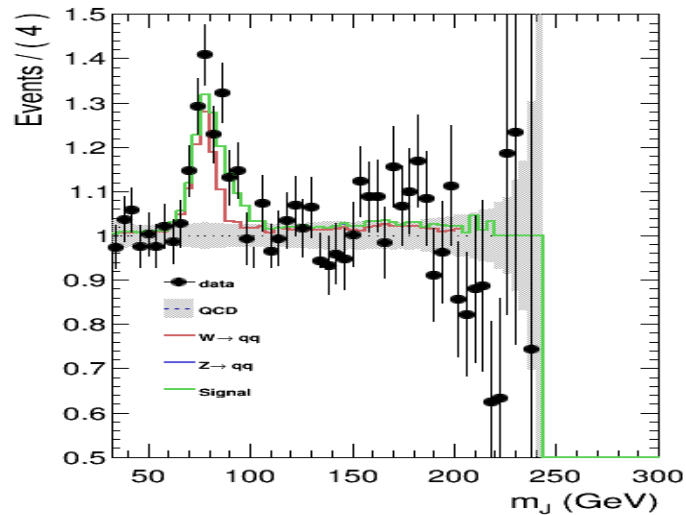
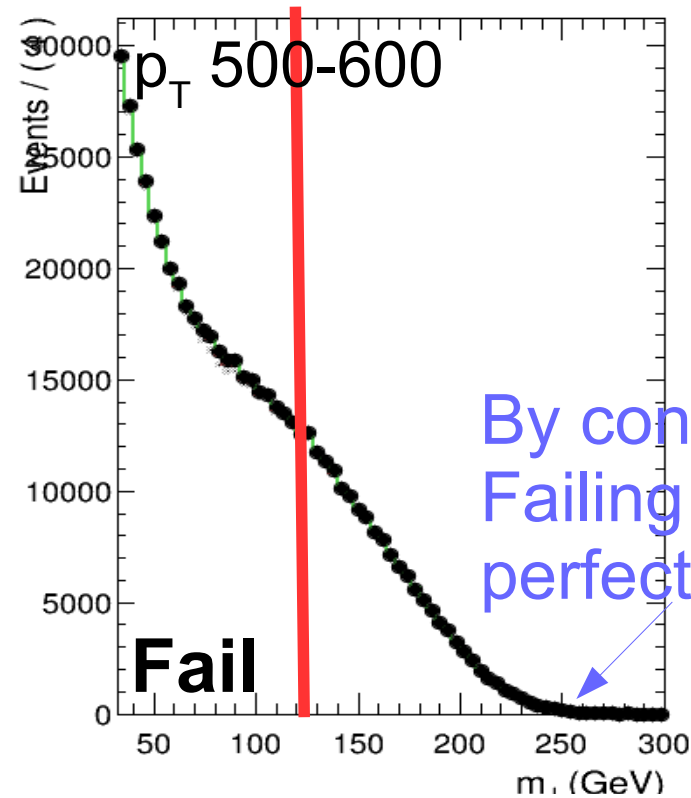
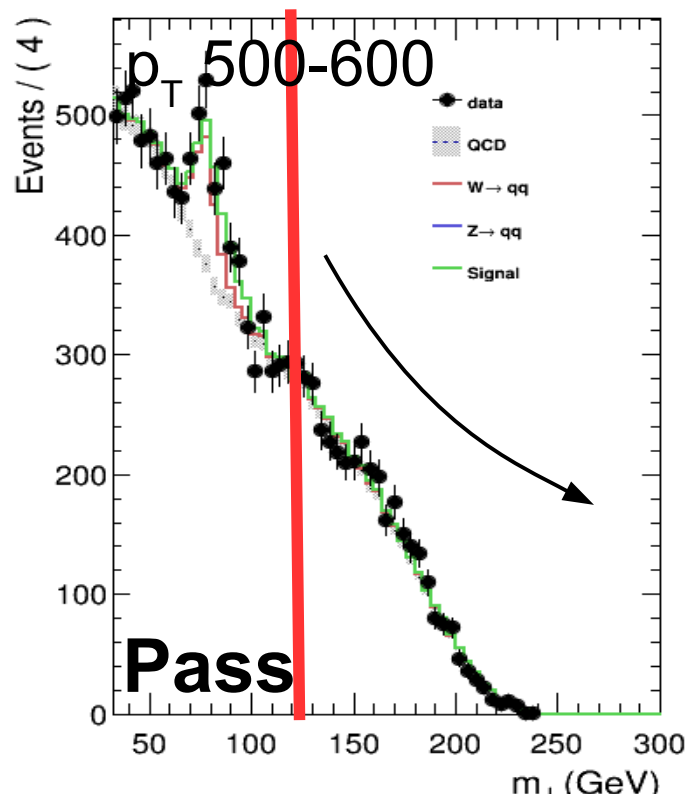
New Directions

- Key concept was modifying jet substructure

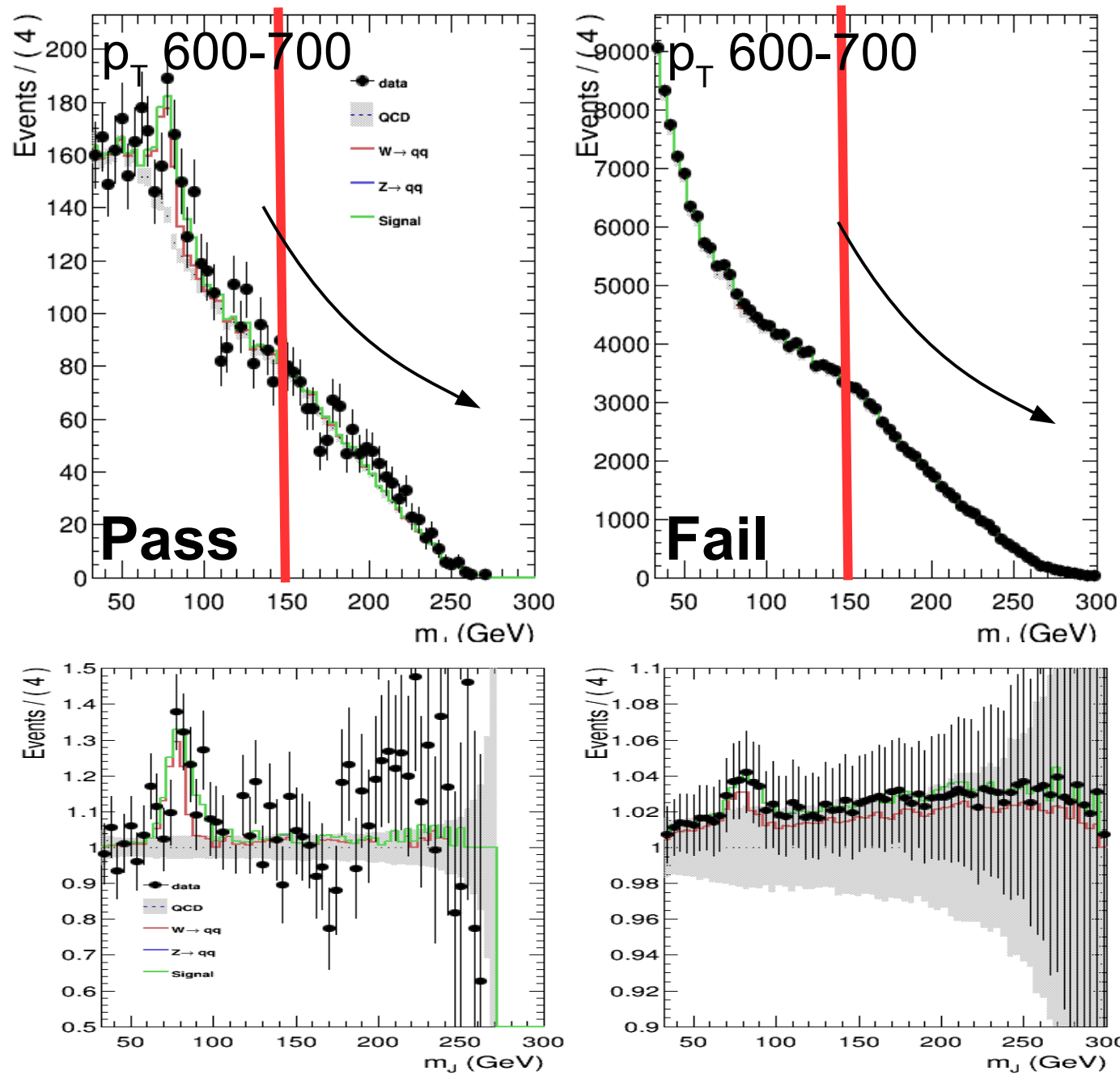


New approaches w/deep learning **can decorrelate observables**

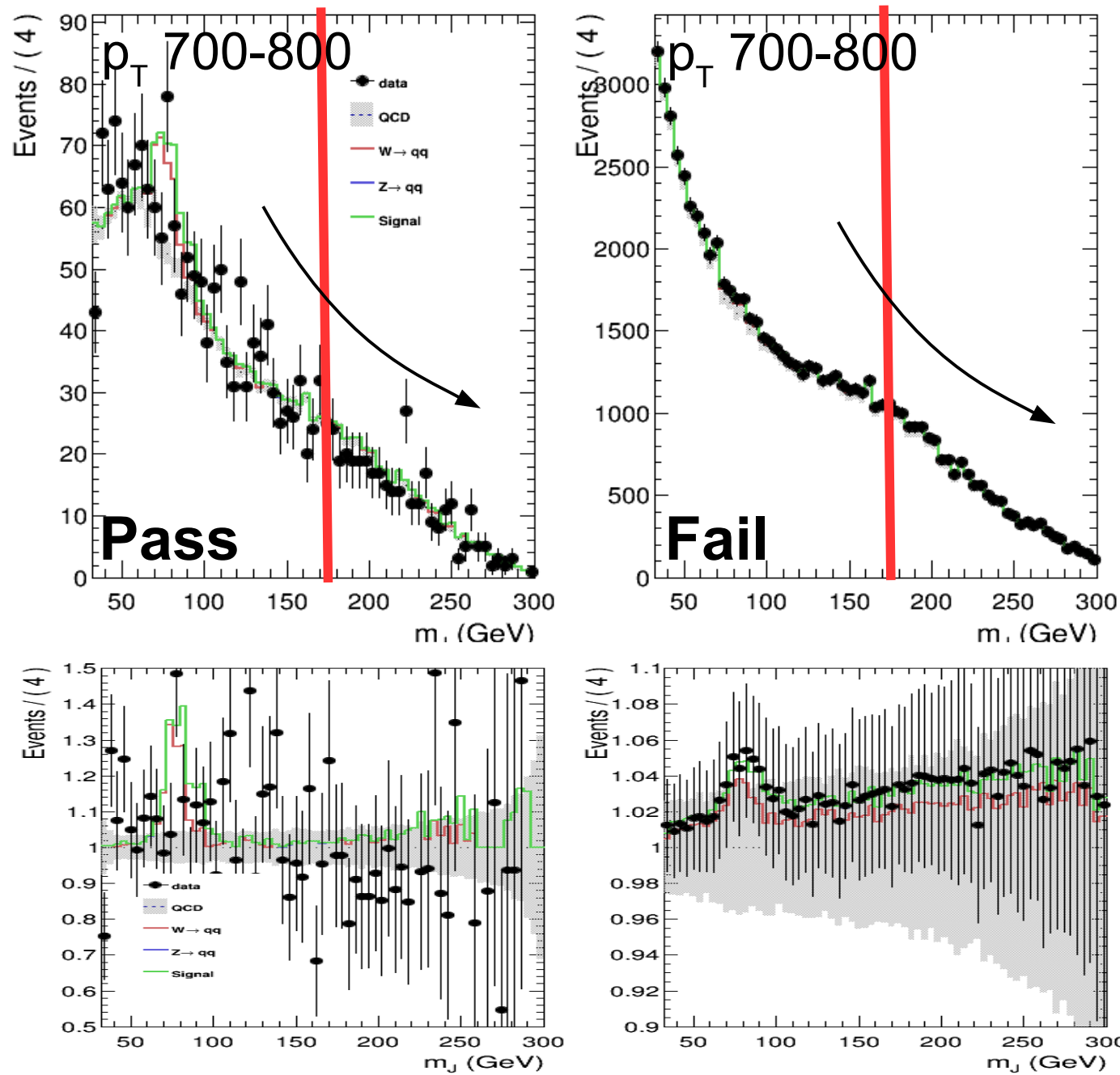
Motion of the Sudakov Peak



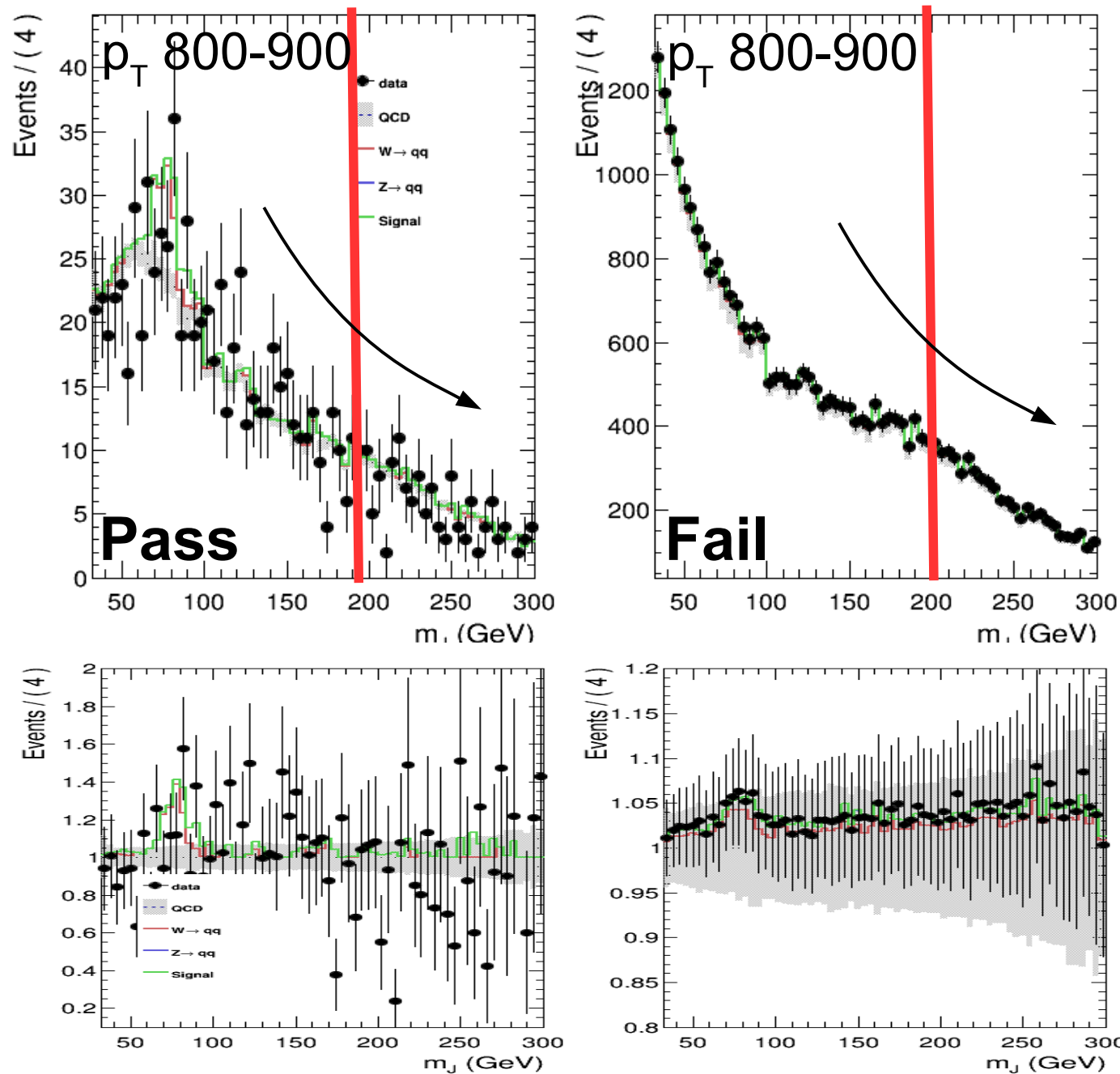
Motion of the Sudakov Peak



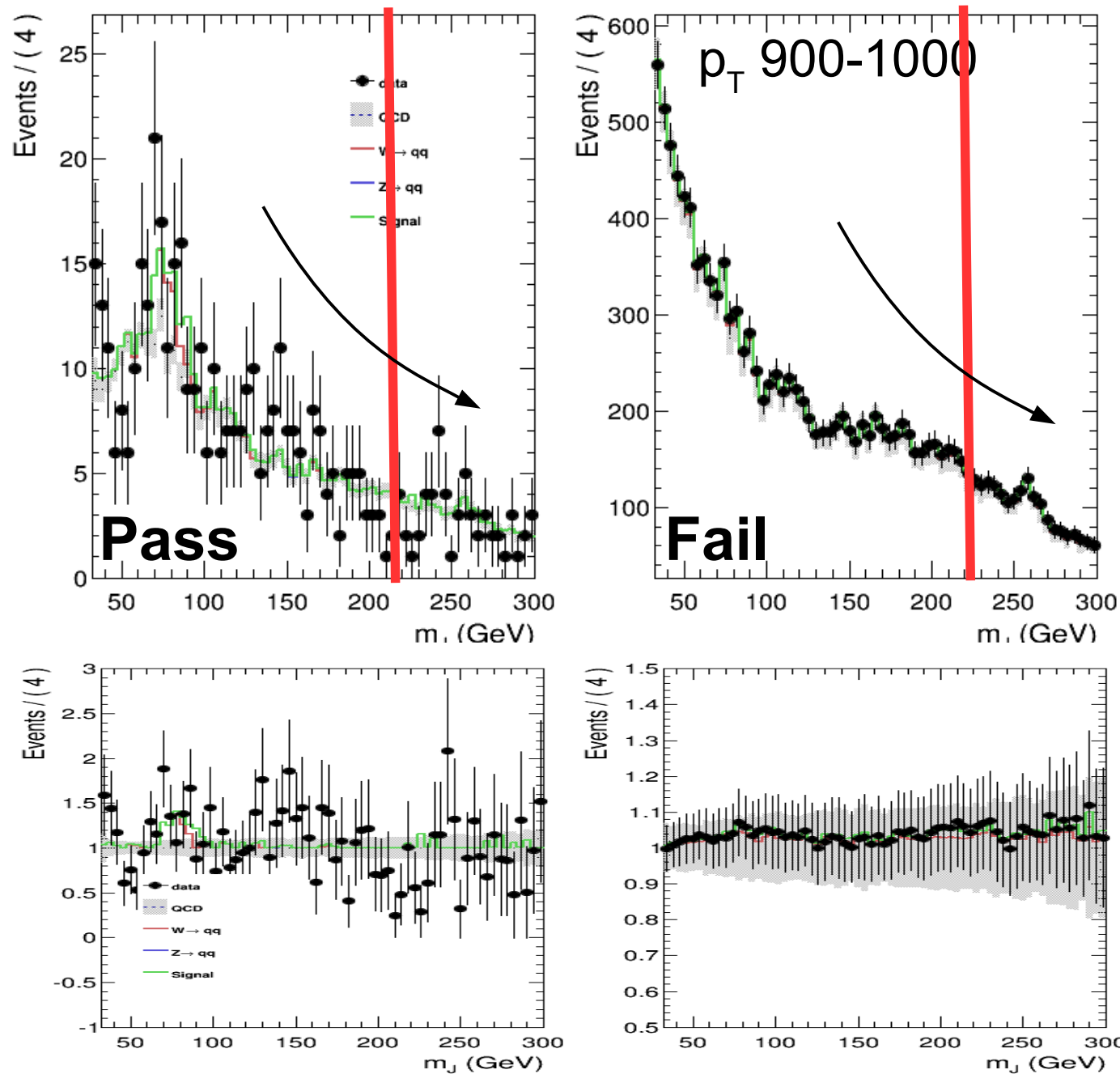
Motion of the Sudakov Peak



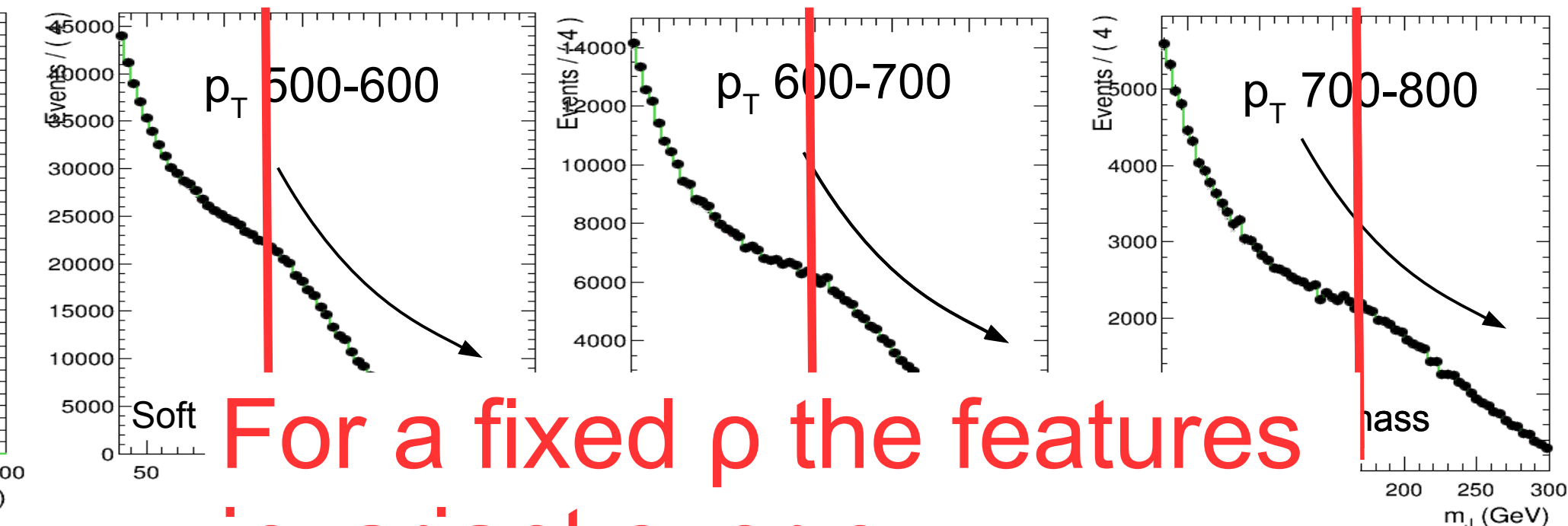
Motion of the Sudakov Peak



Motion of the Sudakov Peak



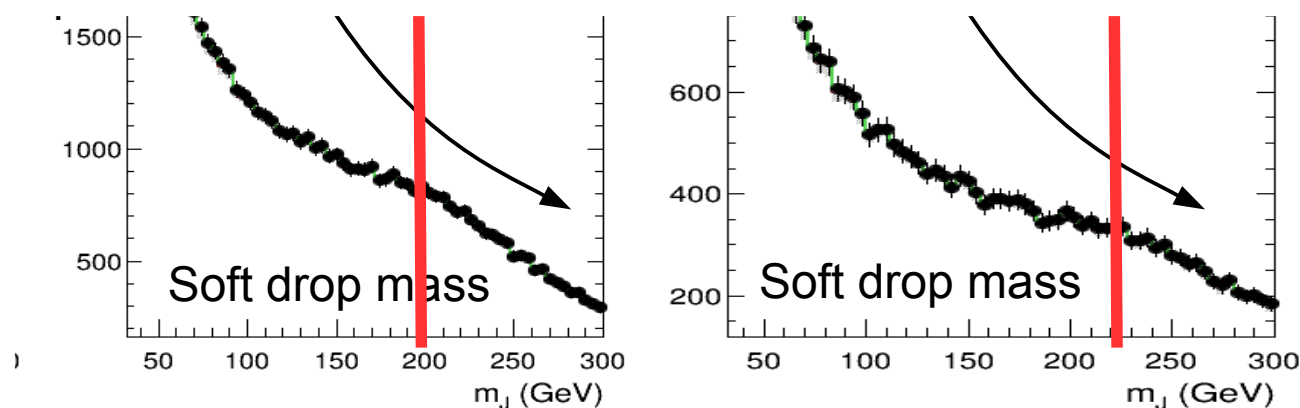
Sudakov peak variation over p_T

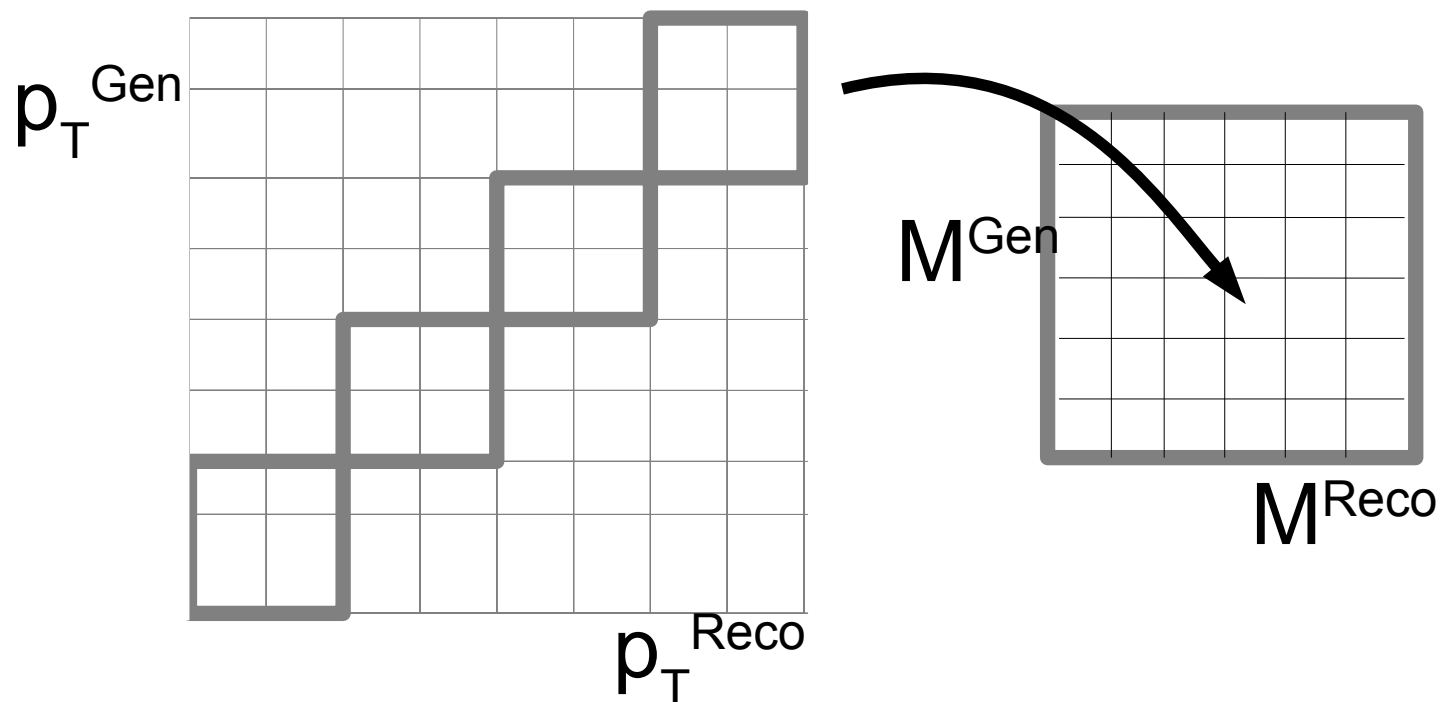


For a fixed p the features
invariant over p_T

Jet Mass
distribut
varies v

n to make
flat
ains
Mass & p_T





Unfolded

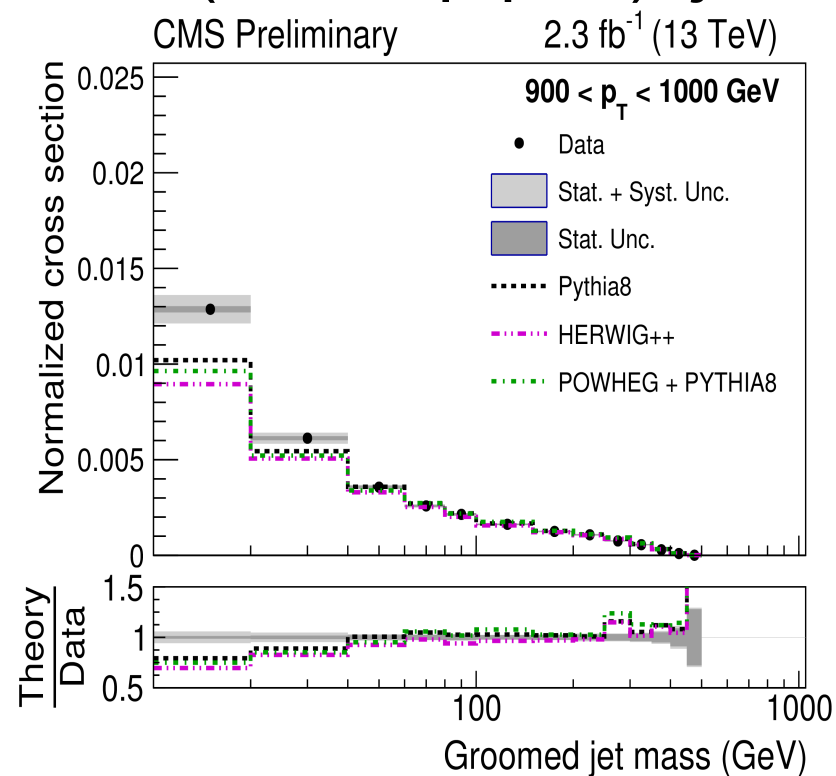
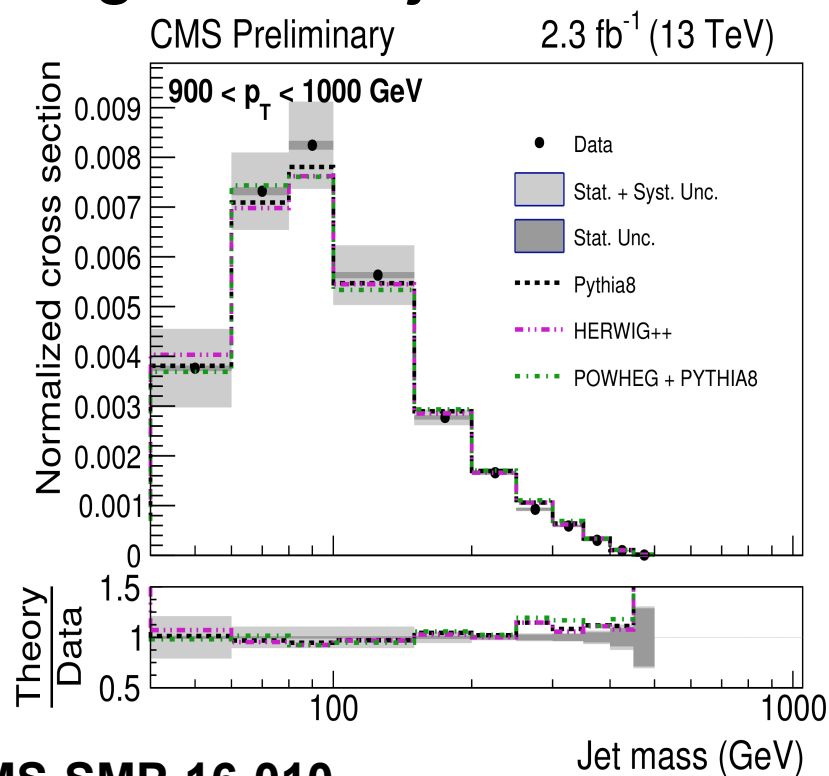
4D unfolding

$M^{\text{reco}}, p_T^{\text{reco}}$ to

$M^{\text{Gen}}, p_T^{\text{Gen}}$

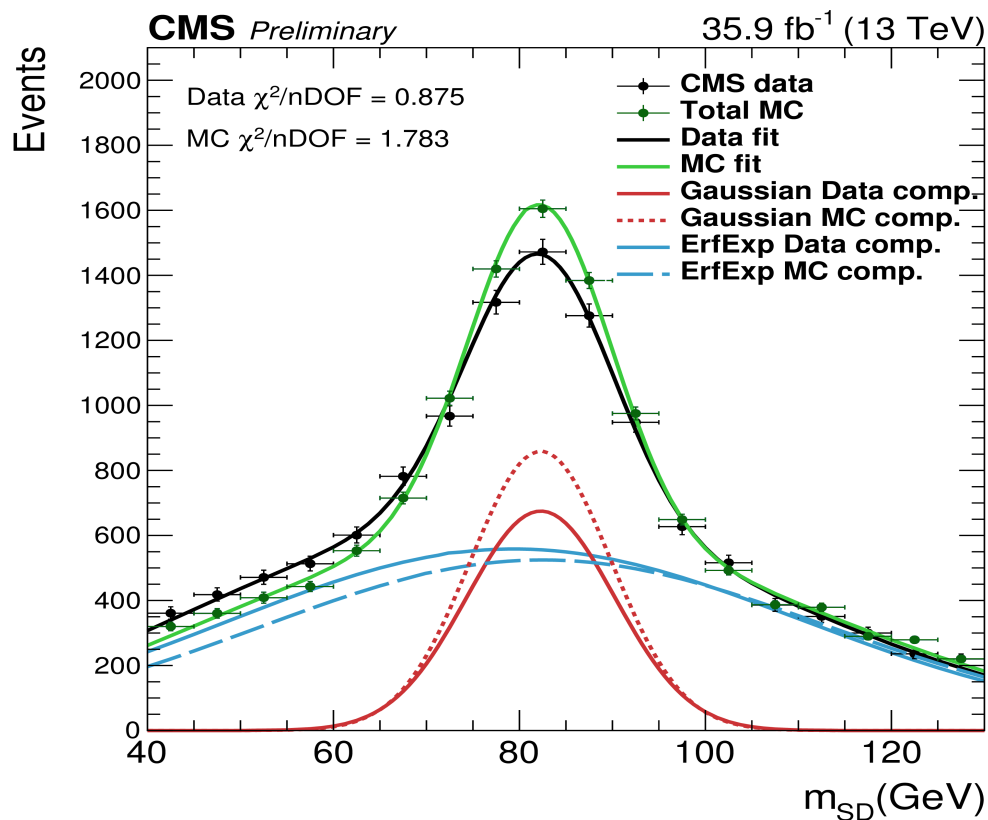
Ungroomed jet mass

mMDT(Sofdrop $\beta=0$) jet mass



Jet Mass Resolution

- Mass resolution obtained with top events

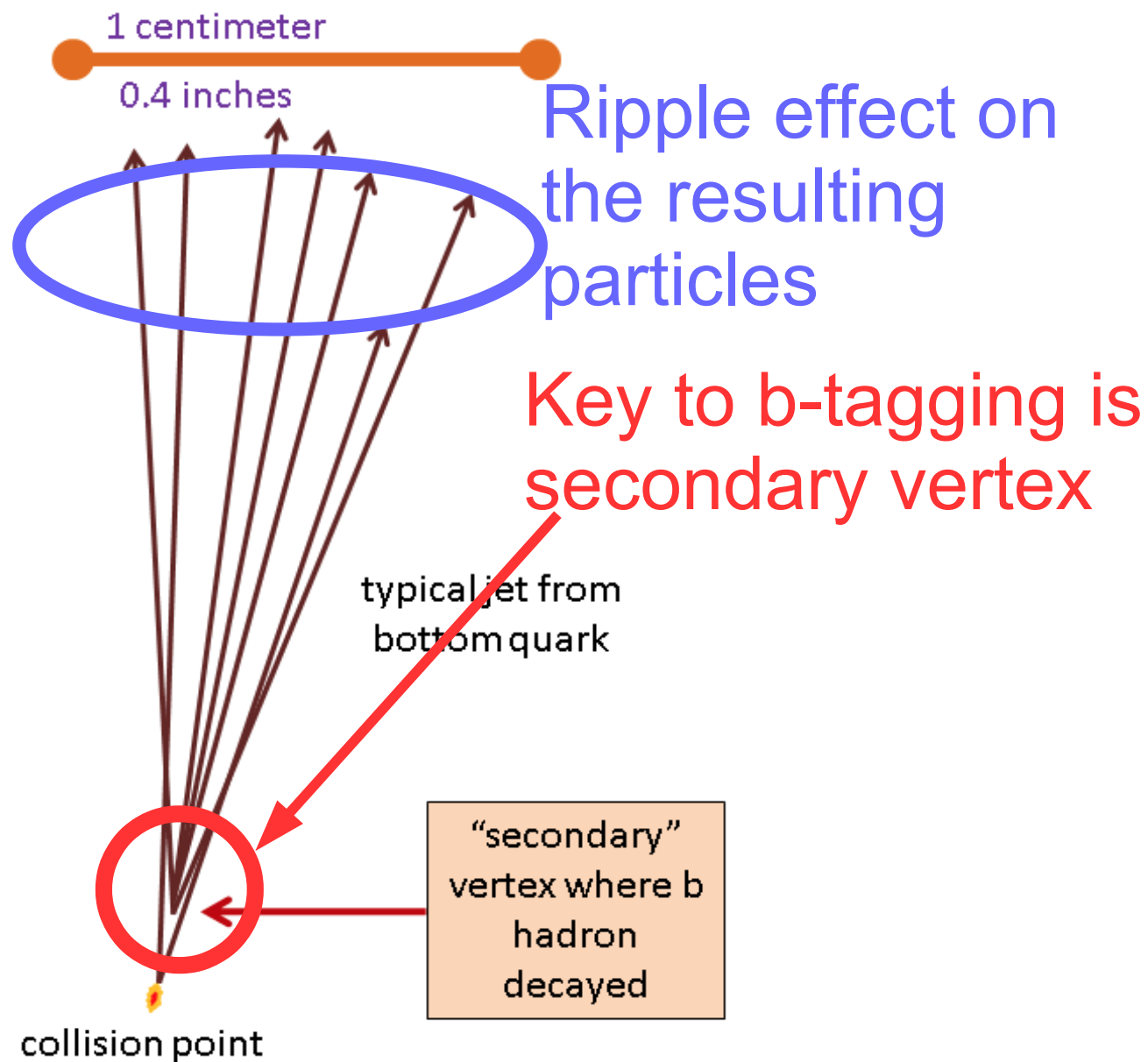
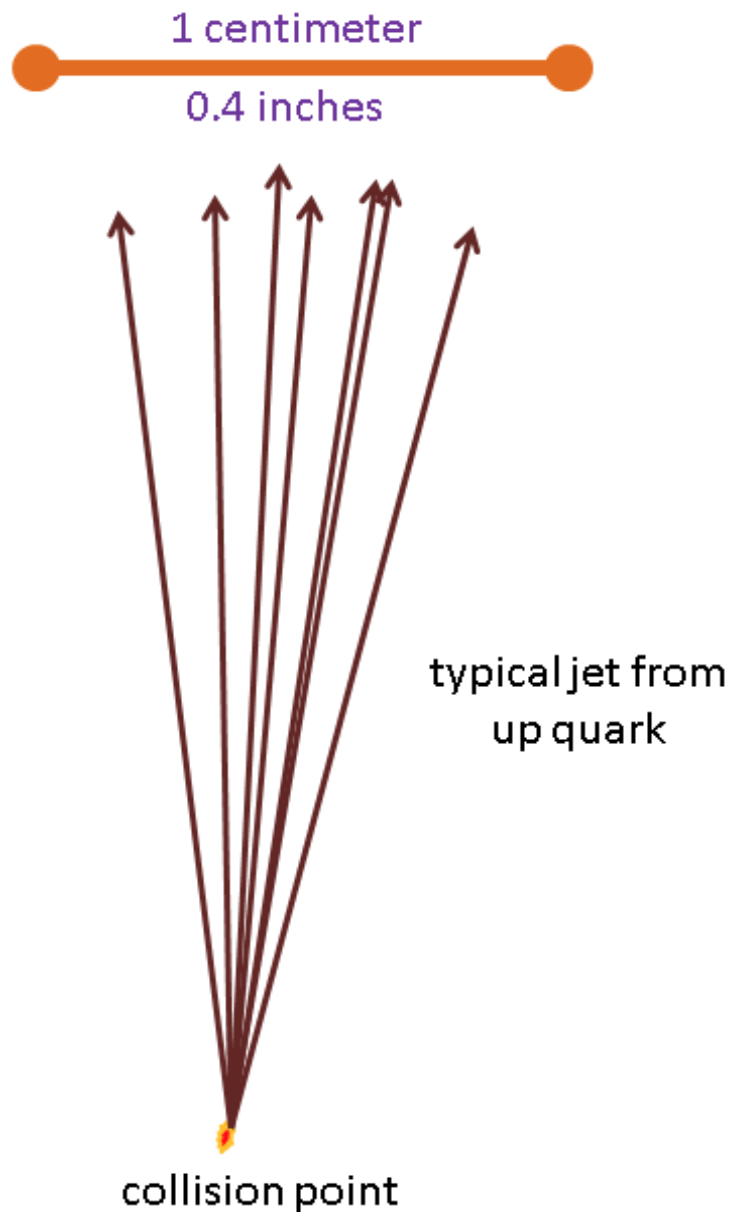


Jet mass and resolution calibrated on W peak in $t\bar{t}$ events

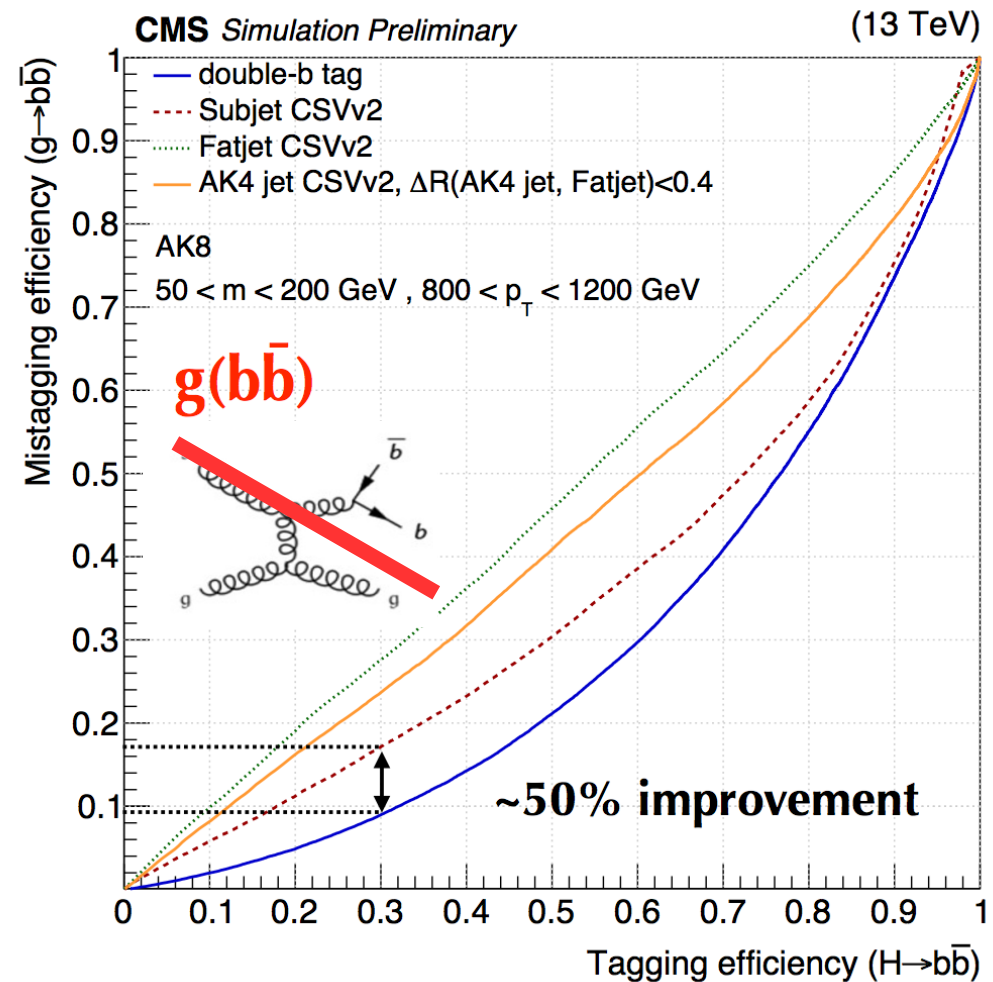
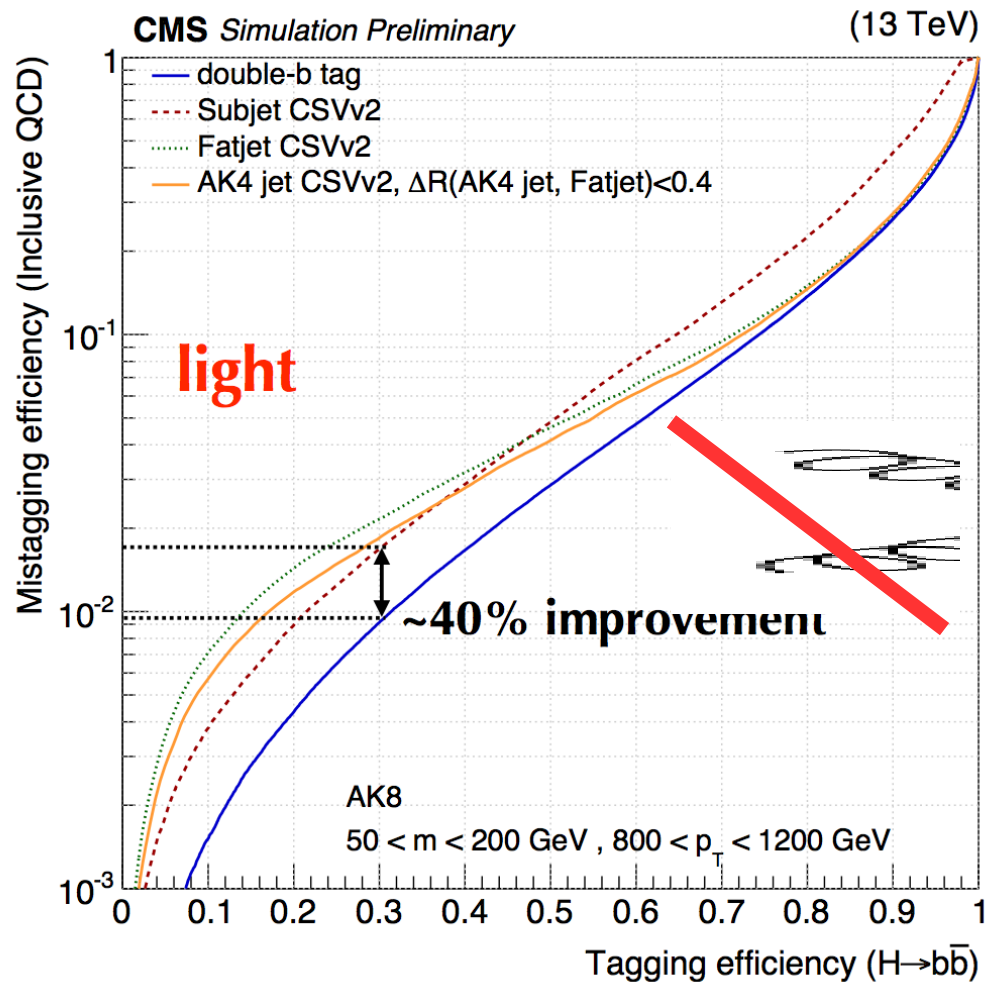
Unc extrapolated to all phase space

In addition compute uncertainty based :
Jet energy corrections/resolution, Pileup,
PDF and Physics model

A b-tag?

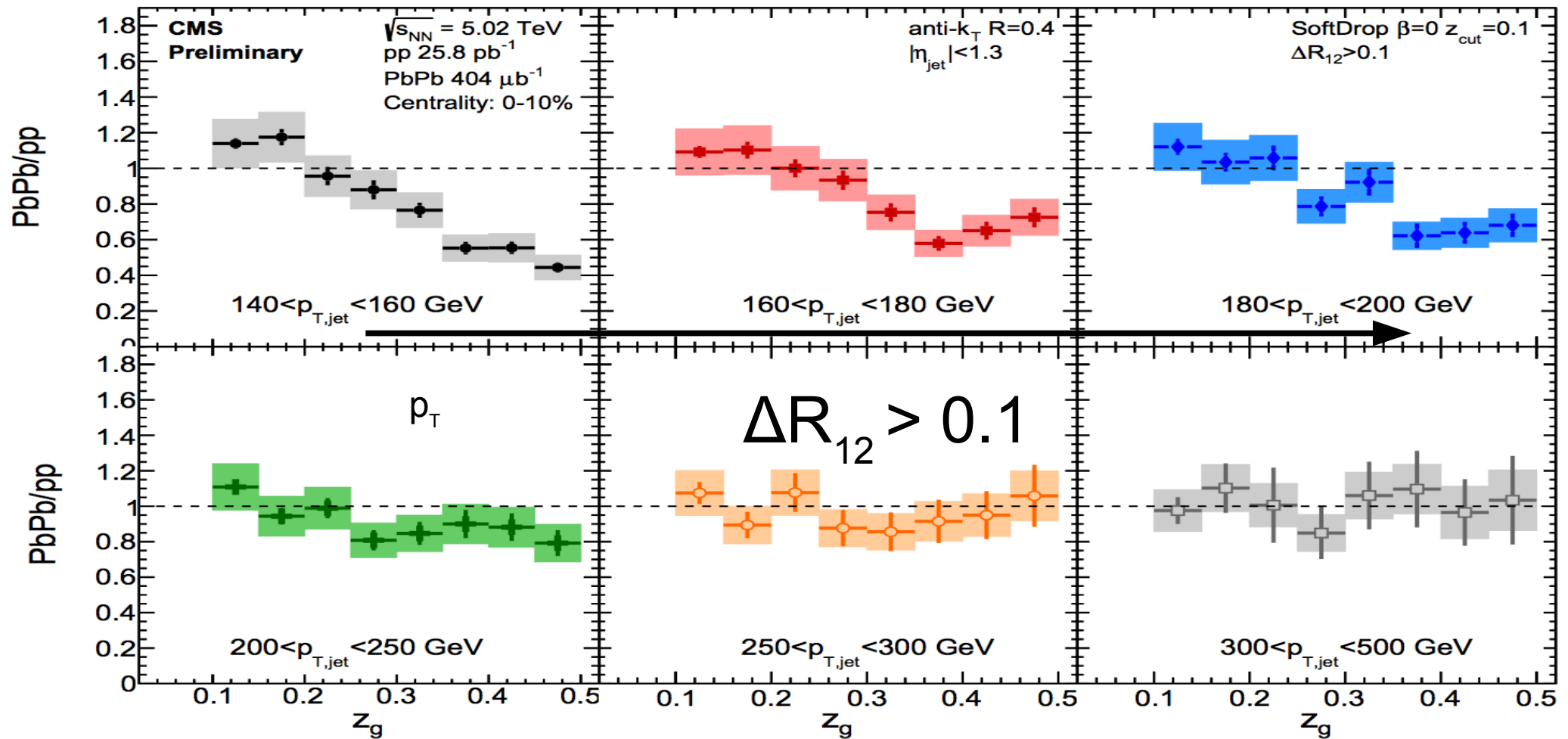


Double B-tagger



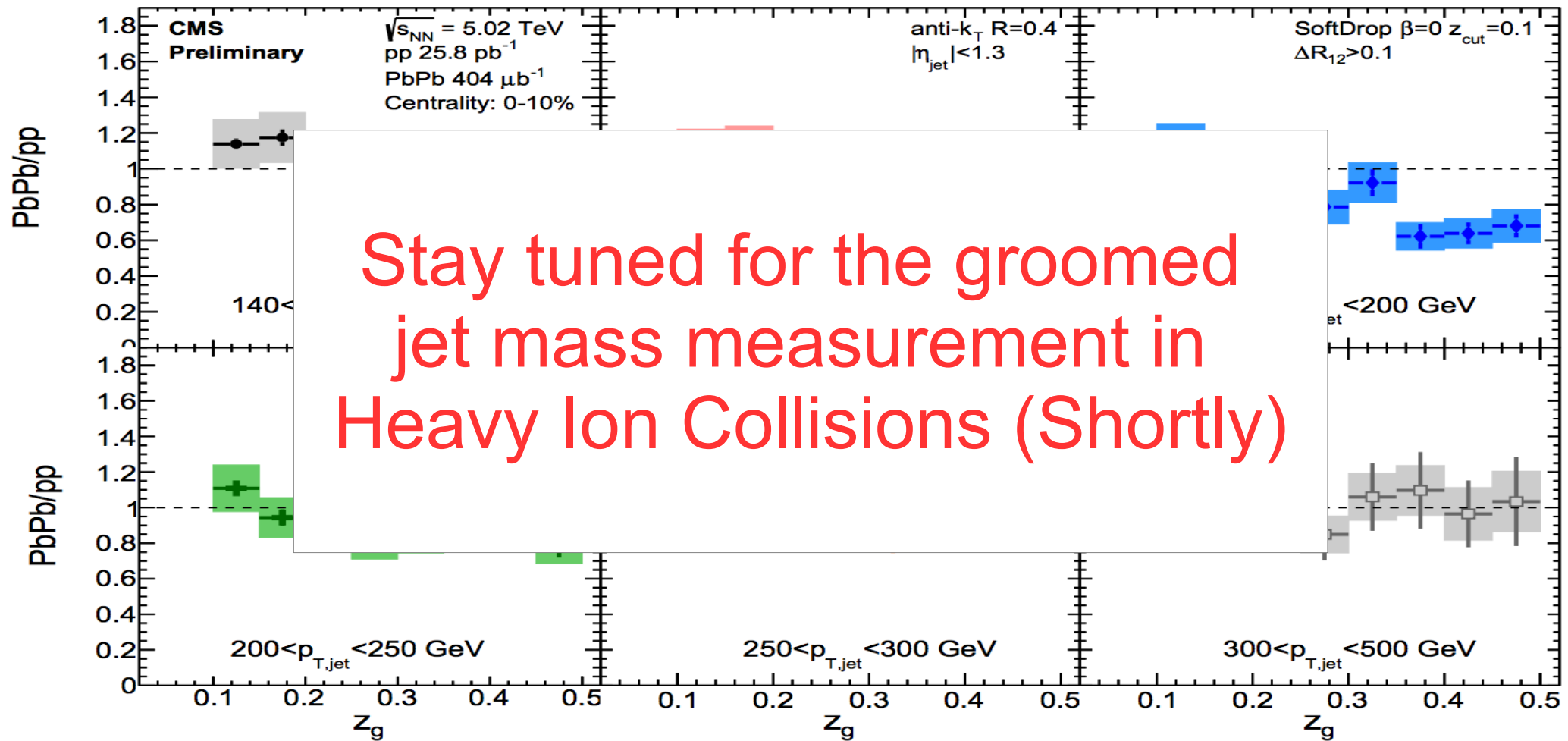
Resulting combination gives 50% improvement over previous

Modification in the medium



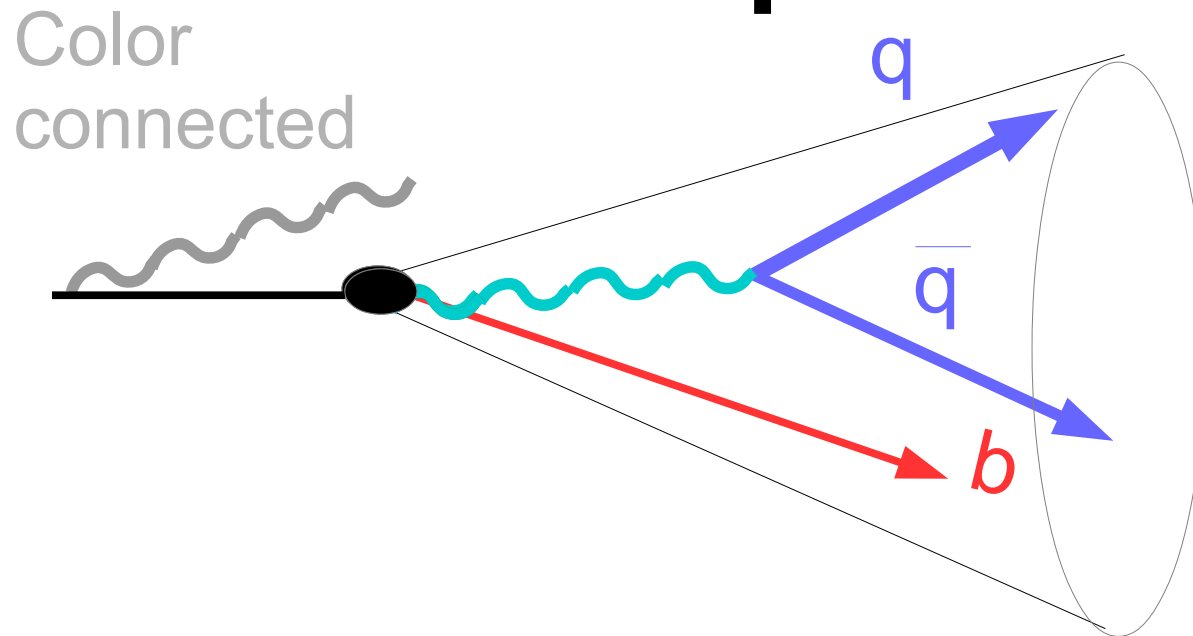
A clear modification is present with subsides at high p_T

Modification in the medium



A clear modification is present with subsides at high p_T

The Top Jet Mass



Top jet decays are a great way to test top mass

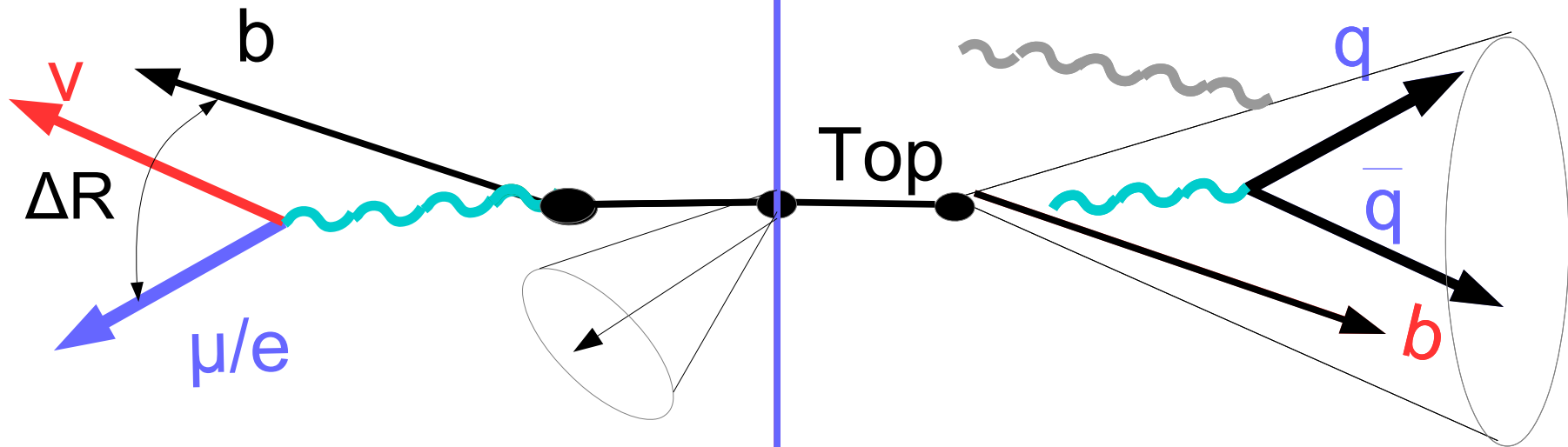
High p_T regime allows us to factorize effects
Potential for a better understanding of mass

Question: How sensitive are we to top mass?

Selecting top events

Name of the game : **don't bias the top mass**

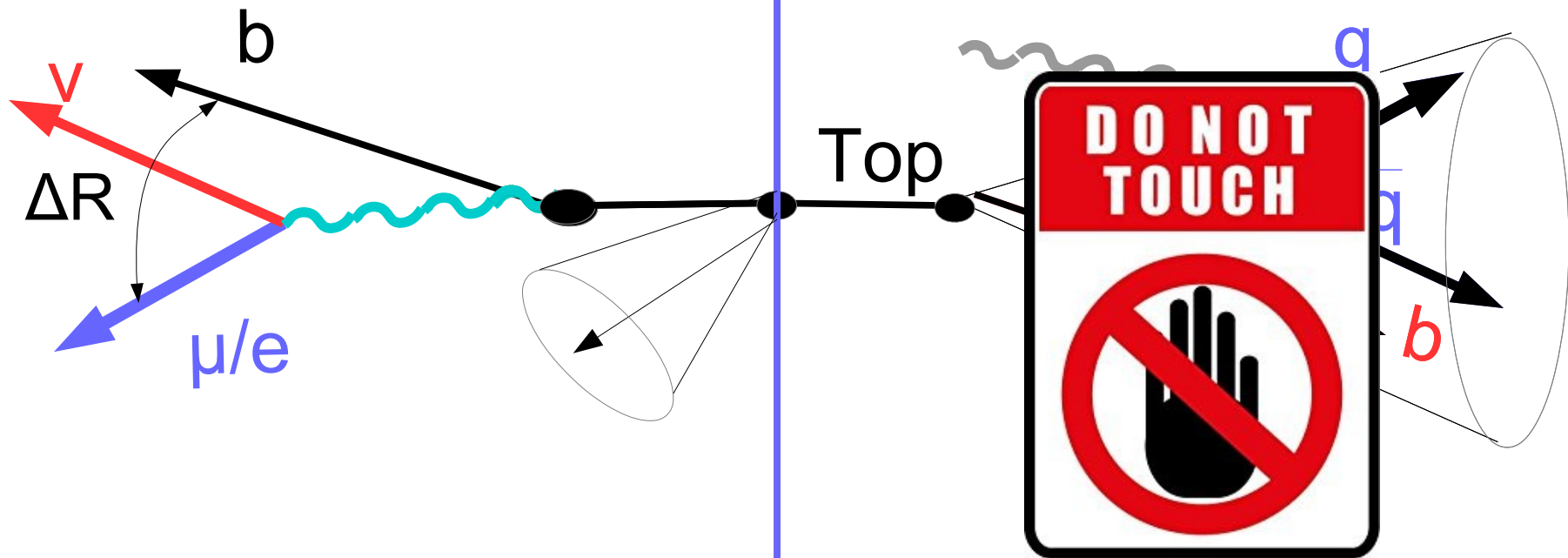
- Goal here is to isolate a pure sample of boosted tops



Selecting top events

Name of the game : **don't bias the top mass**

- Goal here is to isolate a pure sample of boosted tops



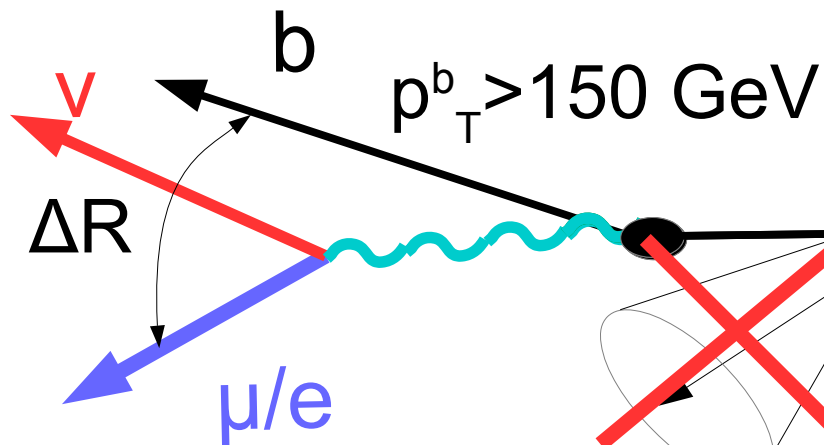
Cut on many parameters

Minimize cut on parameters

Selecting top events

Name of the game : **don't bias the top mass**

- Goal here is to isolate a pure sample of boosted tops



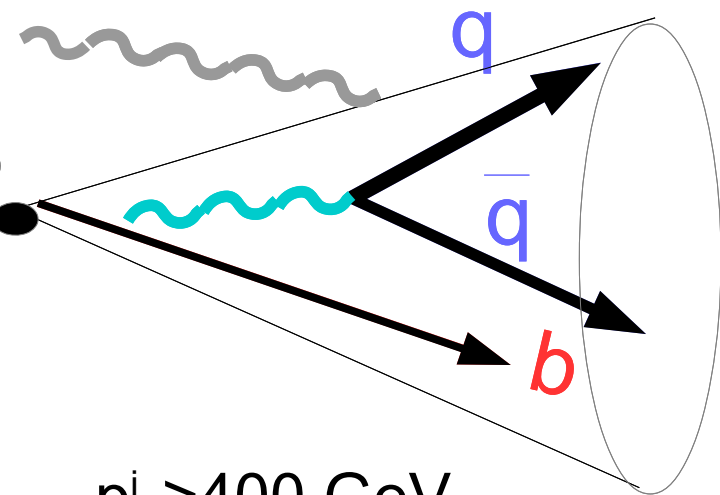
Iso lepton $p_T > 50 \text{ GeV}$

$\Delta R_{lb} < 1.2$

$MET + \text{lepton} > 150 \text{ GeV}$

No additional jet $p_T > 150 \text{ GeV}$

Top



$p_T^j > 400 \text{ GeV}$

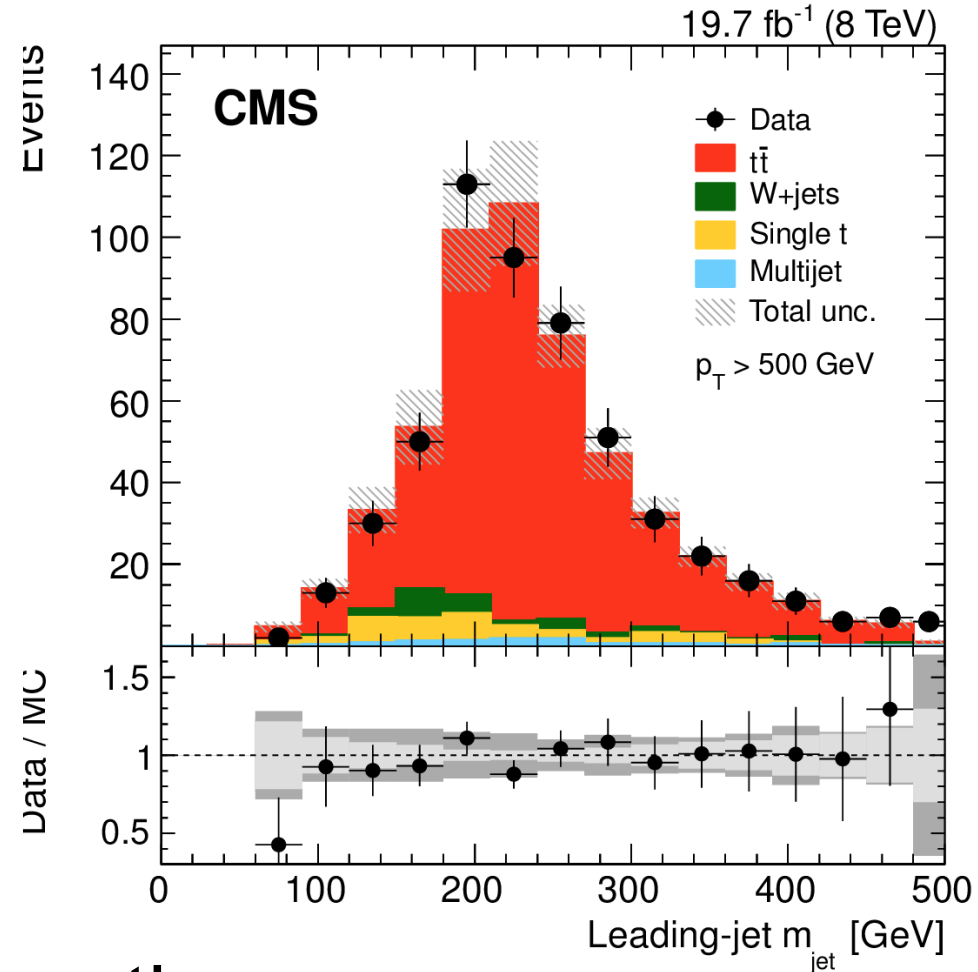
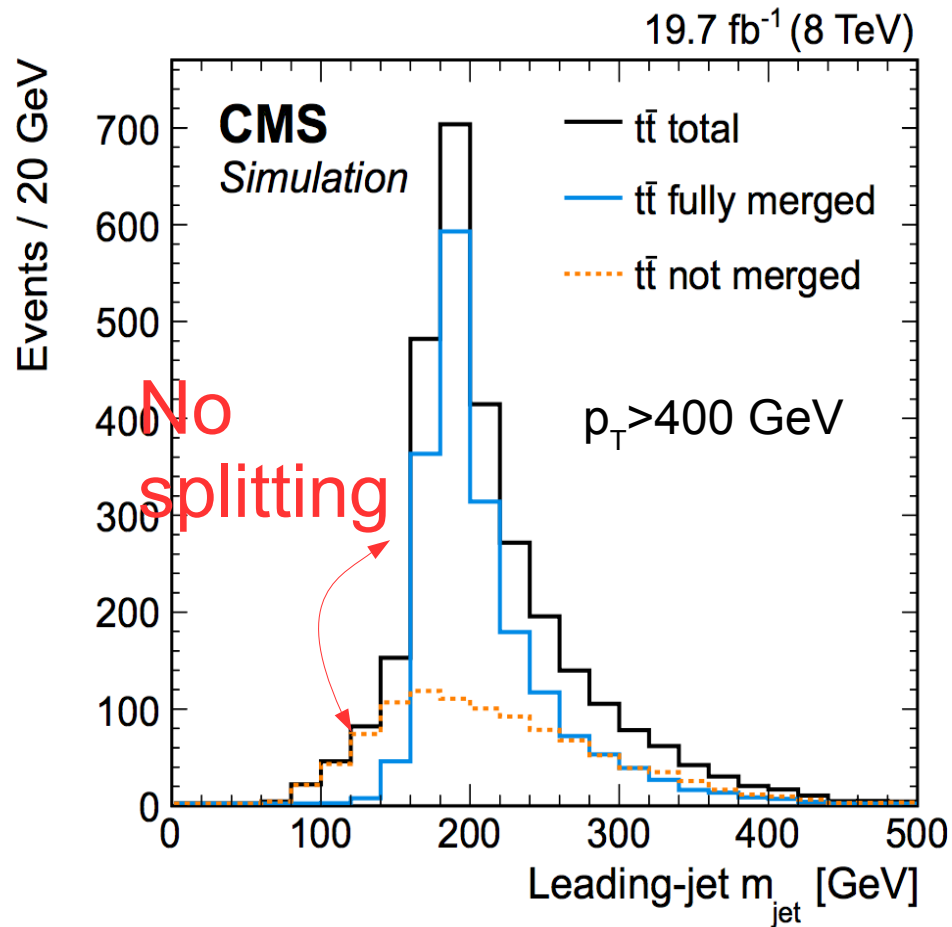
$m^j > m_b^j$

$\Delta R_{\text{cone}} = 1.2$

Cut on many parameters

Minimize cut on parameters

Top Mass

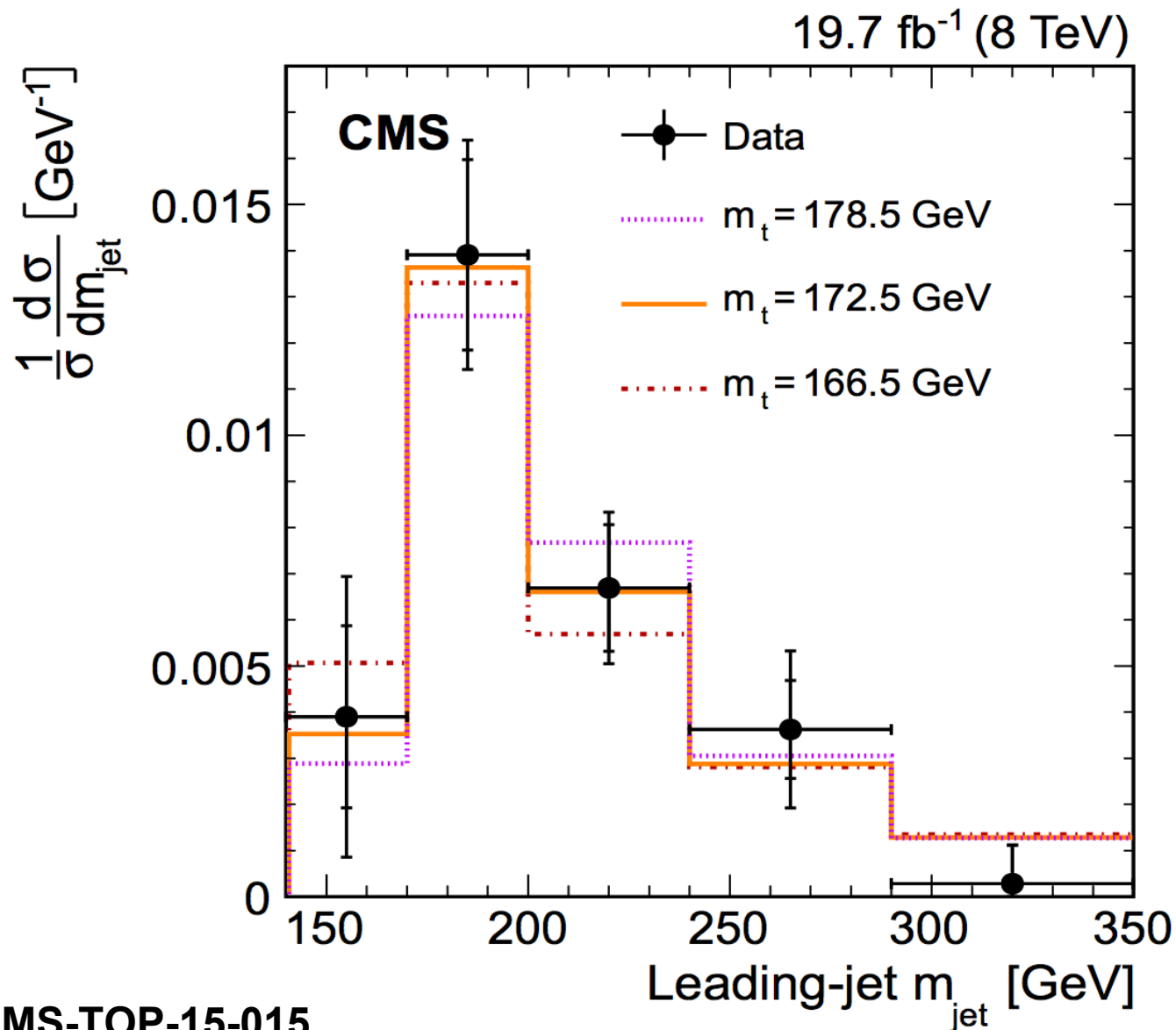


Analysis done inclusive on the mass

Avoid splitting merged and unmerged @ all cost

Goal is **unfold mass model independently**

Top Mass Unfolded

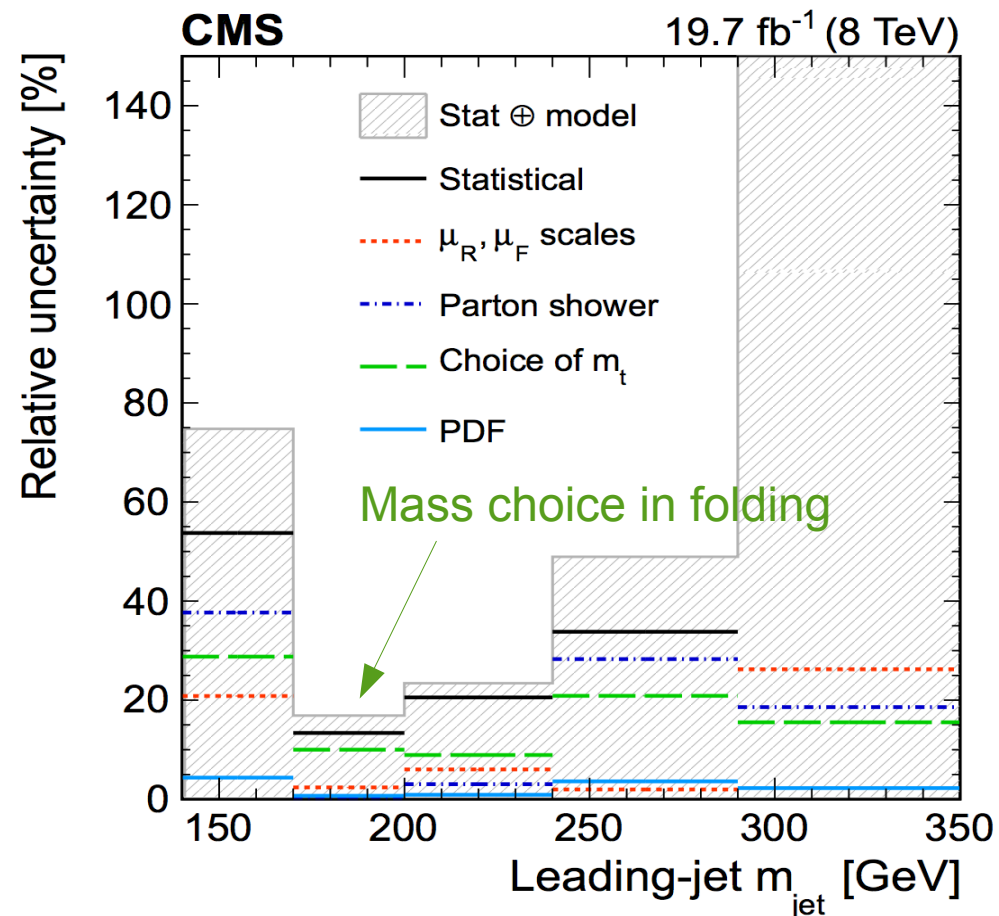
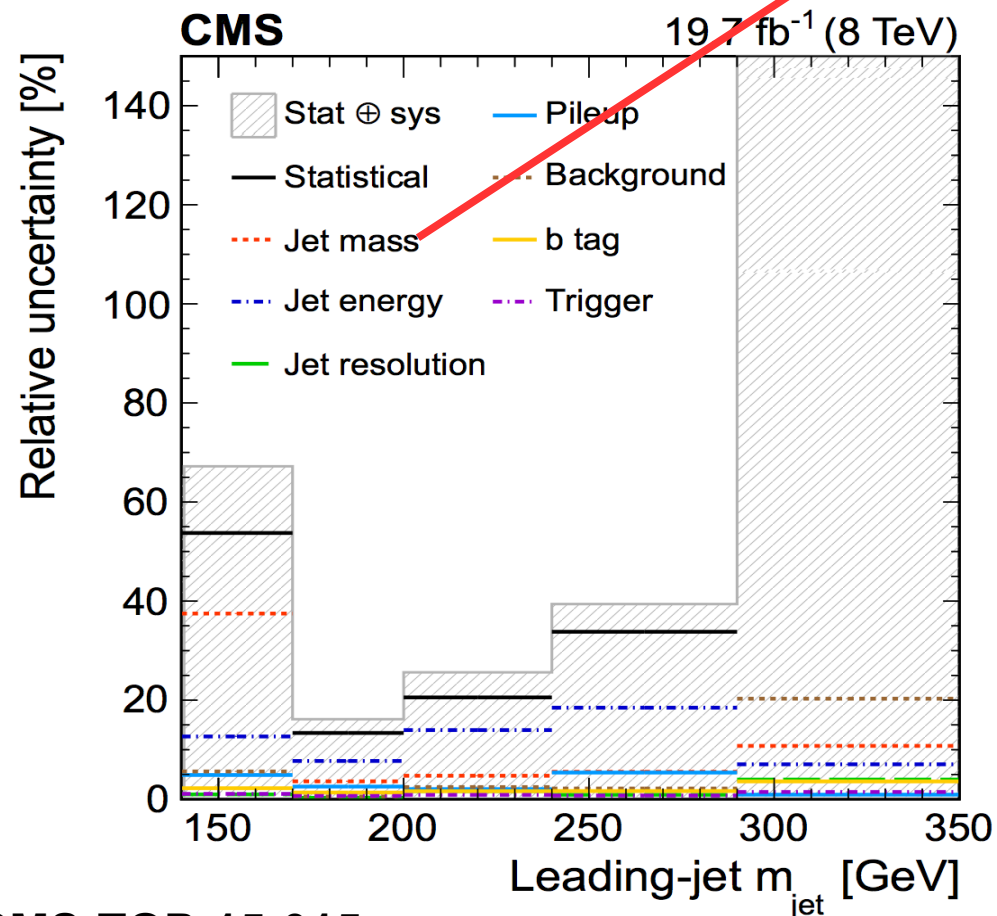
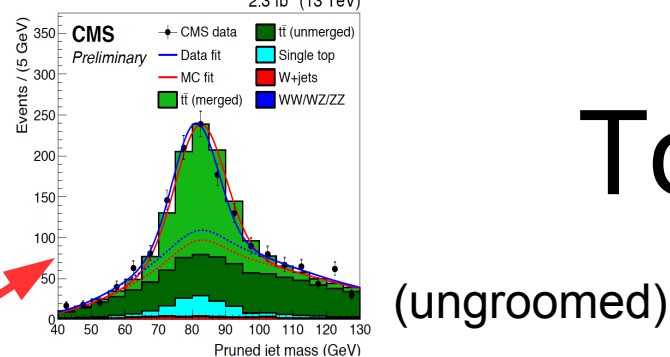


Same scheme as
QCD jet mass

+ accounting for
migration from
tight cuts

Top Mass

- The unfolded result



CMS-TOP-15-015

$$m_t = 170.8 \pm 6.0 \text{ (stat)} \pm 2.8 \text{ (syst)} \pm 4.6 \text{ (model)} \pm 4.0 \text{ (theo)} \text{ GeV}$$

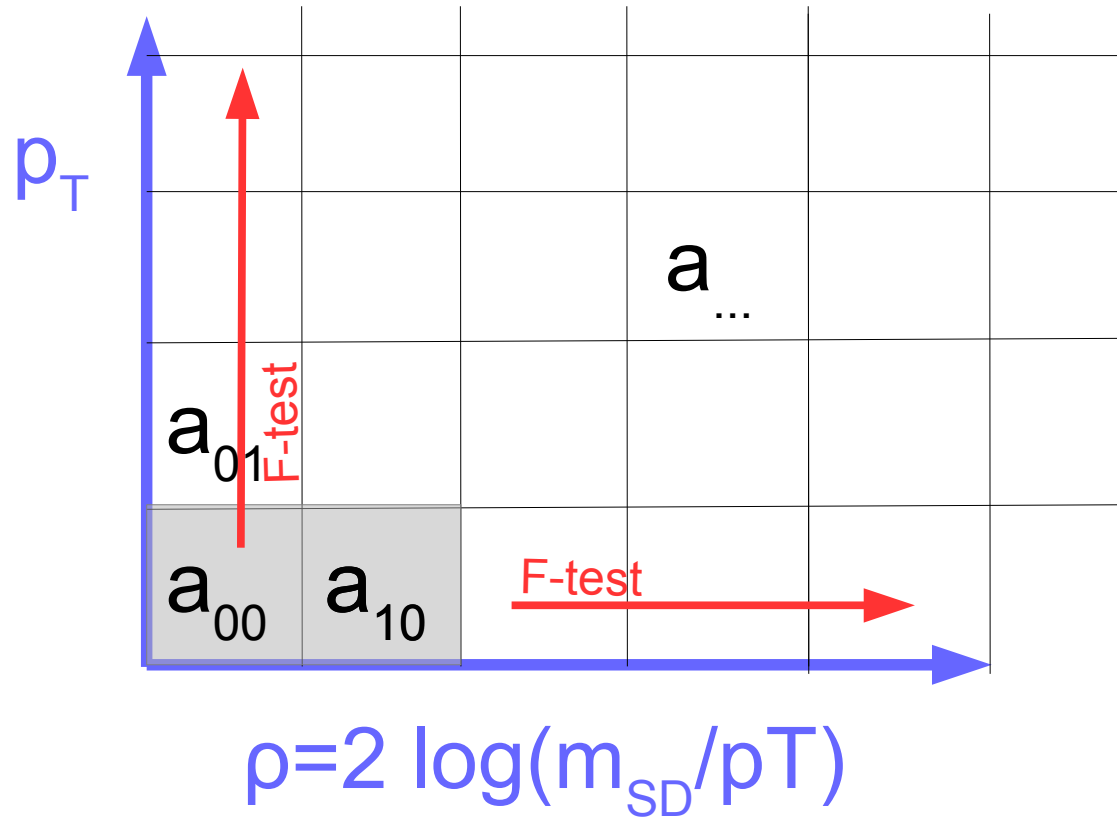
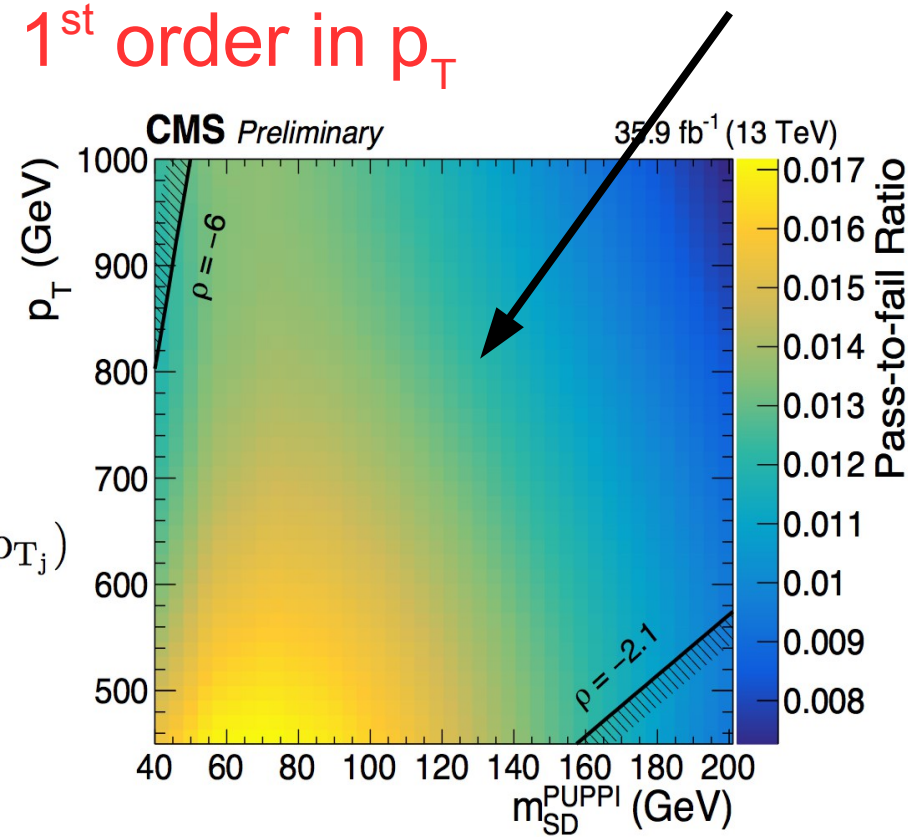
$$= 170.8 \pm 9.0 \text{ GeV,}$$

$$m_t^{\text{SM}} = 173.34 \pm 0.27 \text{ (stat)} \pm 0.71 \text{ (syst)} \text{ GeV}$$

For H peak
 2nd order in ρ
 1st order in p_T

Fitted Surface

In data



Following

$$N_{\text{pass}}^{\text{QCD}}(m_{\text{SD}i}, p_{\text{T}j}) = \epsilon^{\text{QCD}} \cdot \left(\sum_{k,\ell} a_{k\ell} \rho_{ij}^k p_{\text{T}j}^\ell \right) \cdot N_{\text{fail}}^{\text{QCD}}(m_{\text{SD}i}, p_{\text{T}j}) .$$