

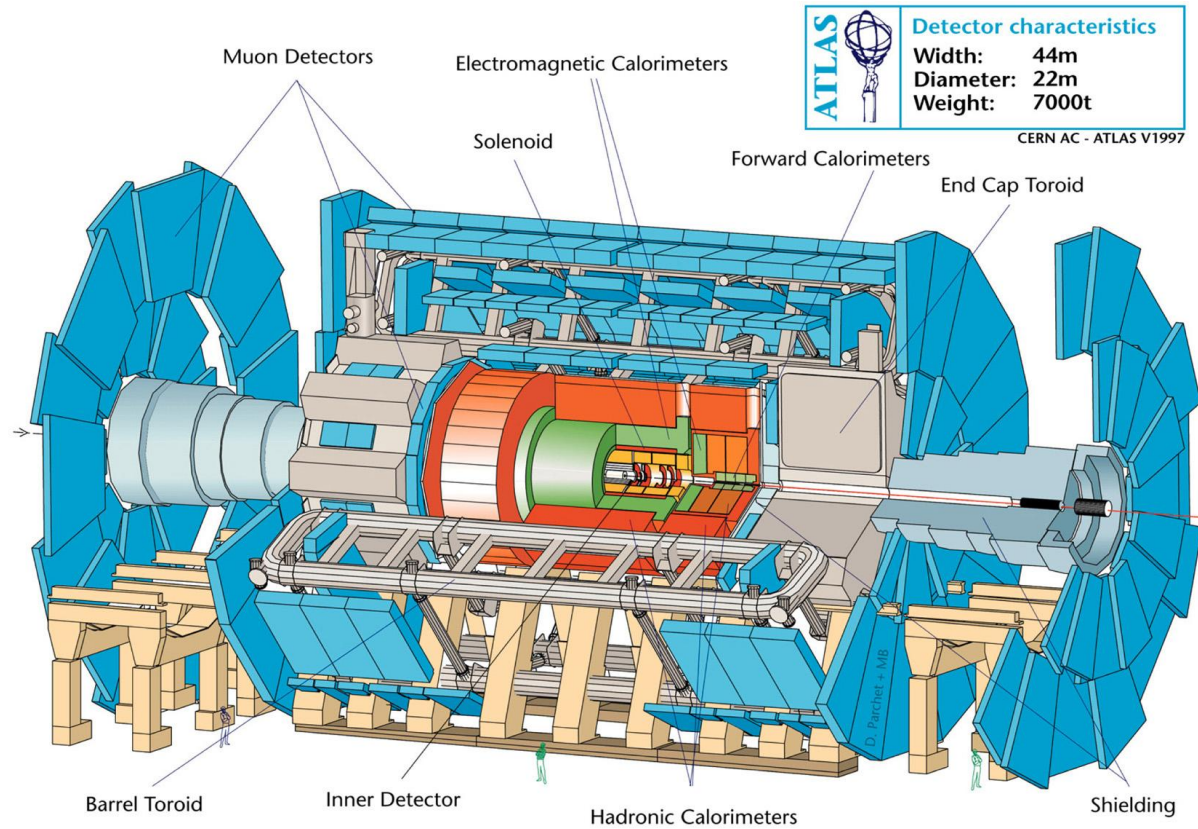
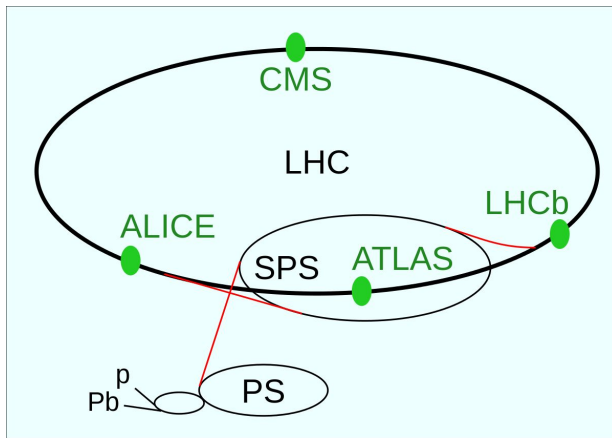
# Classifying Quark and Gluon Jets for ATLAS at CERN

Gareth Smith

Supervisor: Dag Gillberg

September 17th 2018  
Carleton Student Presentations

# The ATLAS experiment

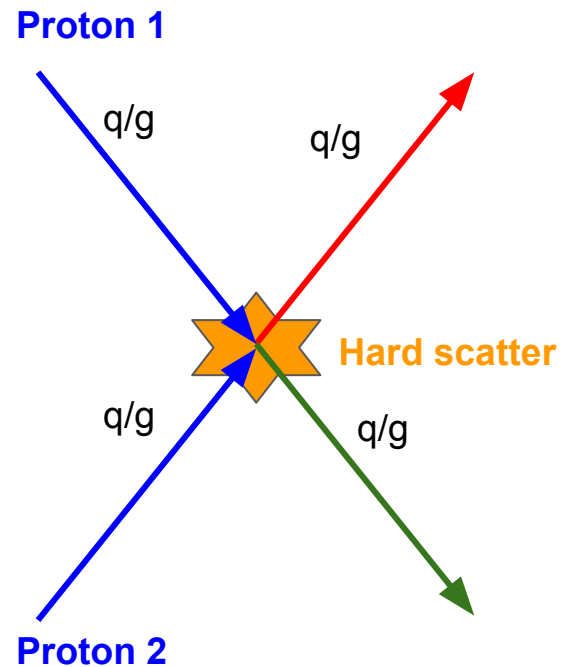


- The ATLAS experiment is a general purpose experiment at the LHC.
- It is used to study proton-proton collisions at 13 TeV.

# What is a jet?

---

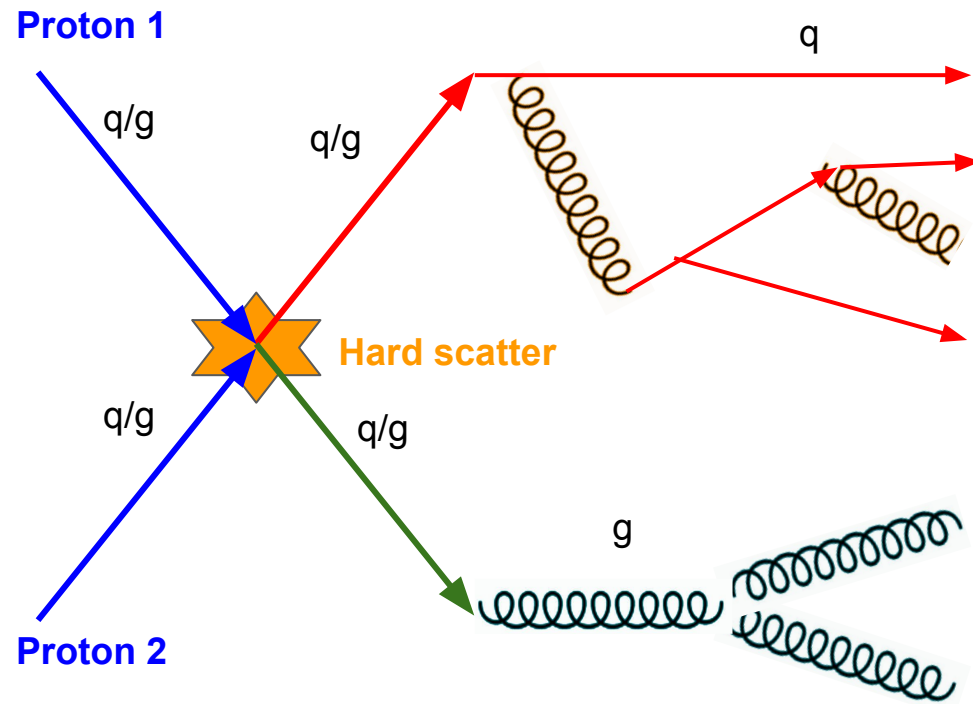
# What is a jet?



Matrix element level

Collision between  
two partons in LHC.  
Results in two  
partons at first order.

# What is a jet?



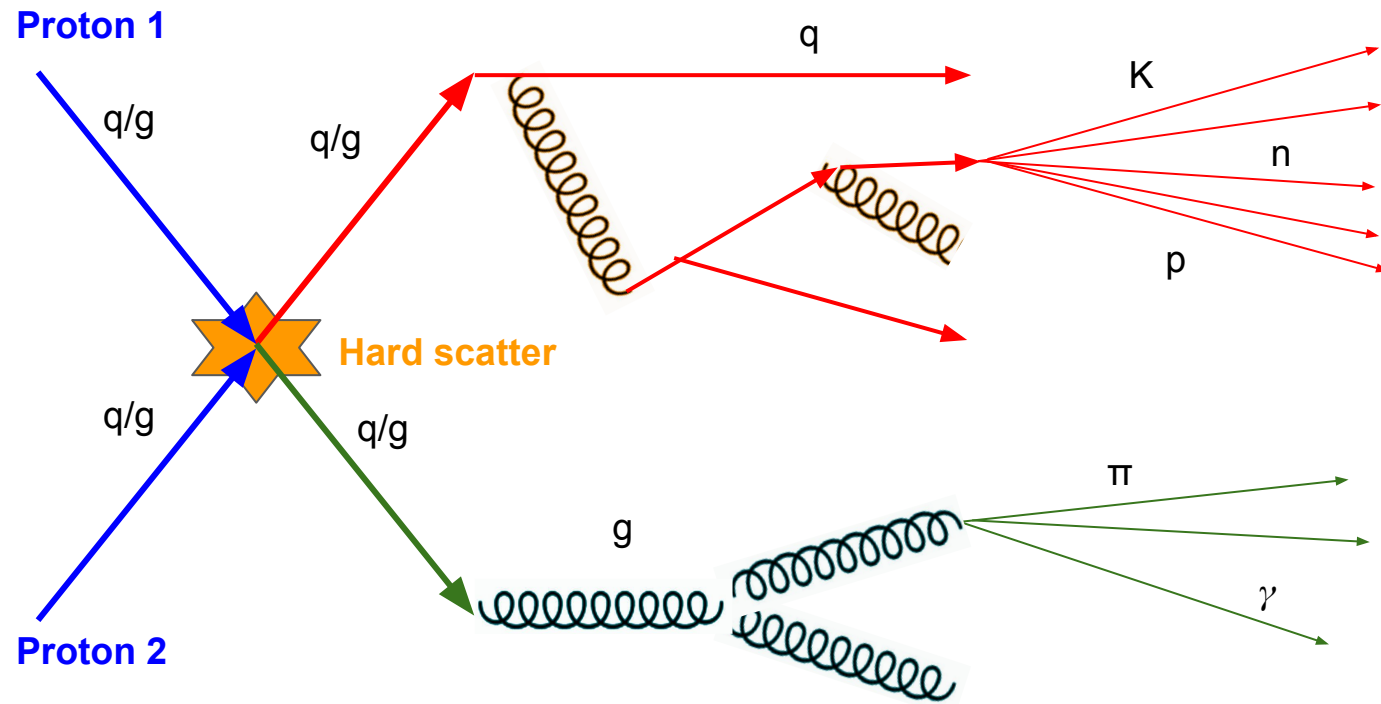
## Matrix element level

Collision between two partons in LHC. Results in two partons at first order.

## Parton shower level

Partons undergo QCD interactions and create "parton jets".

# What is a jet?



## Matrix element level

Collision between two partons in LHC. Results in two partons at first order.

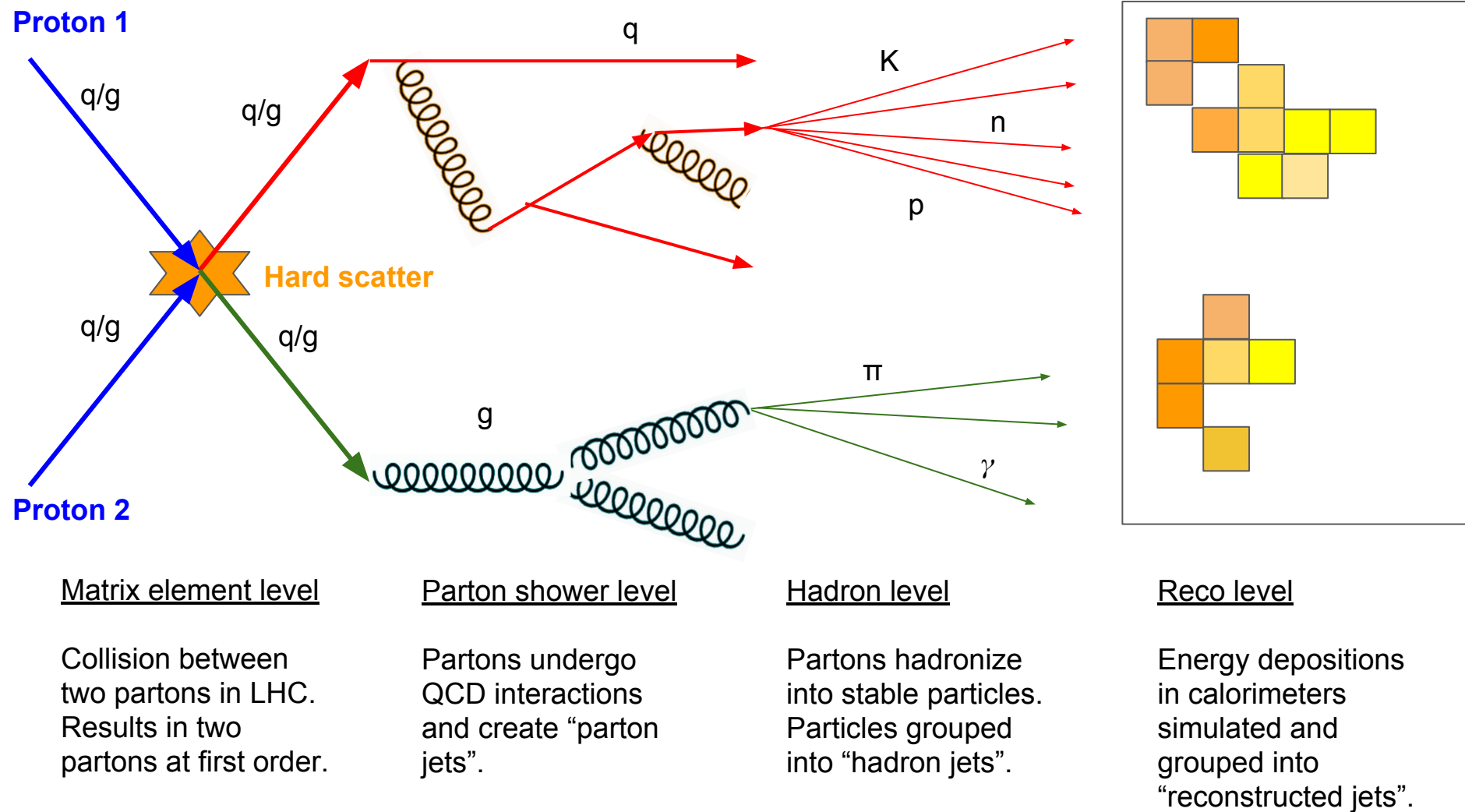
## Parton shower level

Partons undergo QCD interactions and create "parton jets".

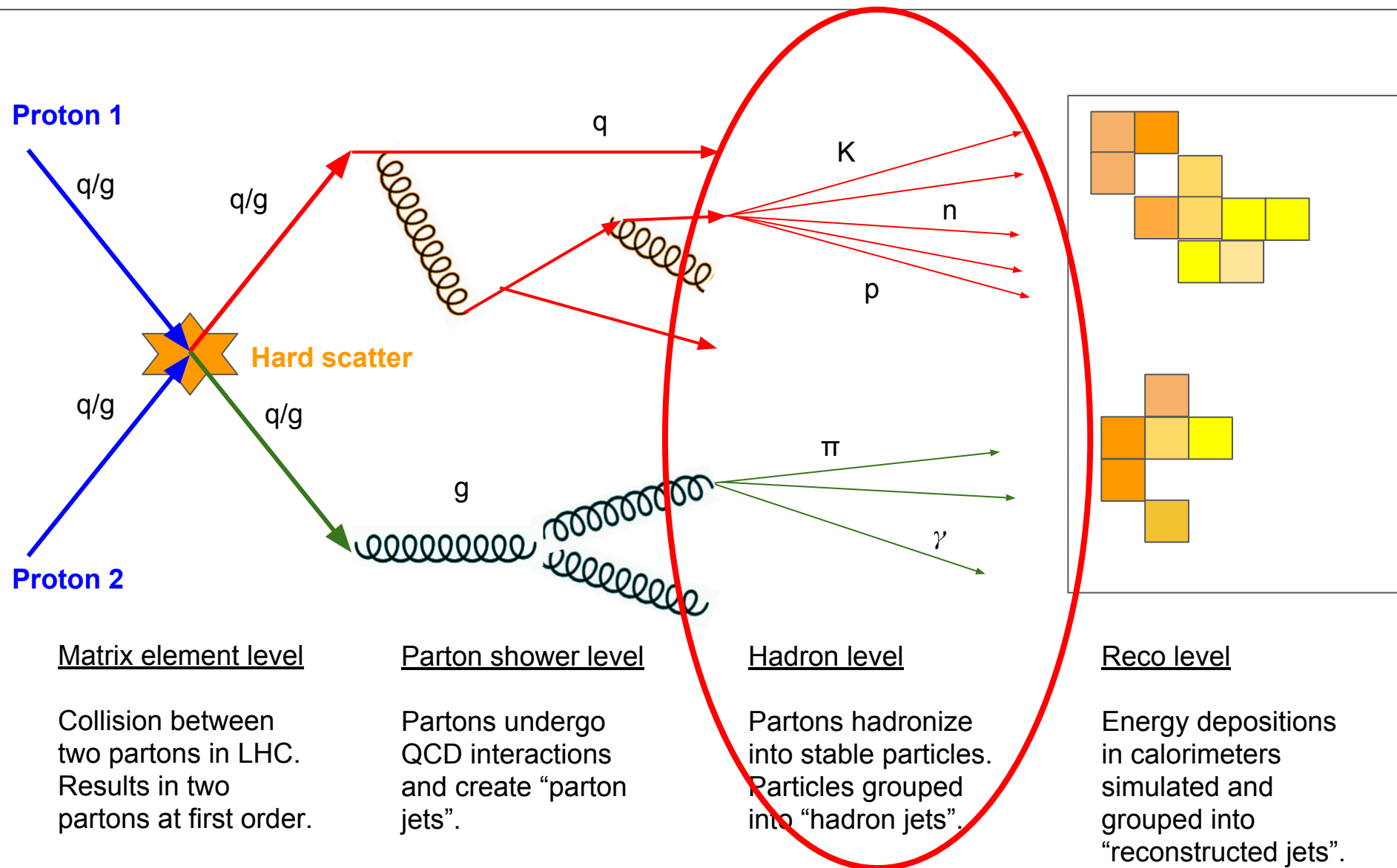
## Hadron level

Partons hadronize into stable particles. Particles grouped into "hadron jets".

# What is a jet?



# What is a jet?



## Matrix element level

Collision between two partons in LHC. Results in two partons at first order.

## Parton shower level

Partons undergo QCD interactions and create "parton jets".

## Hadron level

Partons hadronize into stable particles. Particles grouped into "hadron jets".

## Reco level

Energy depositions in calorimeters simulated and grouped into "reconstructed jets".



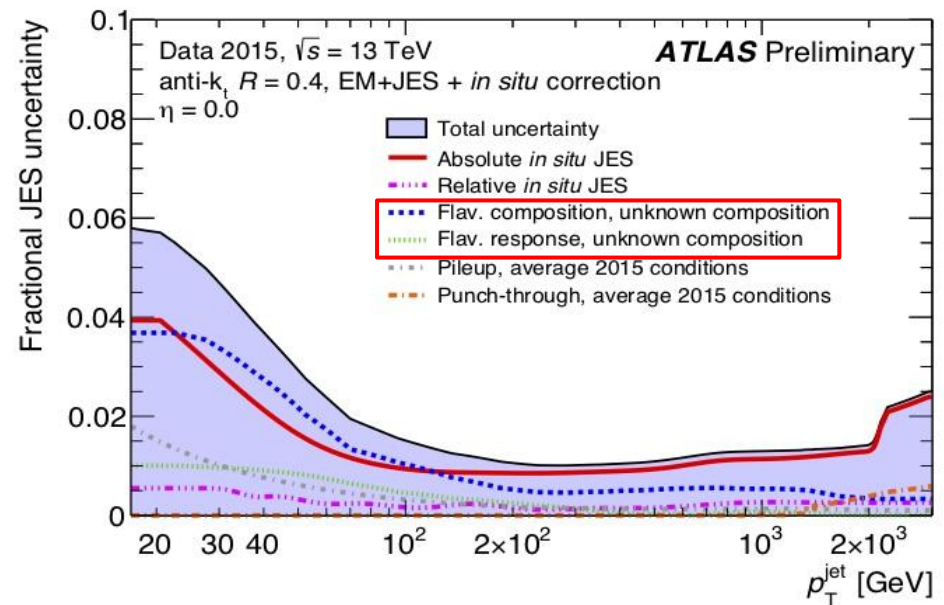
# Why do we want a jet classifier?

- Interested in calorimeter **response**, aka Jet Energy Scale (**JES**).

Ratio of transverse momentum at reco level and hadron level.

$$R = p_T^{\text{reco}} / p_T^{\text{hadron}}$$

- Response depends on flavour of jet (initiated by quark or gluon).
- For many analyses, *Jet Energy Scale uncertainty* is a dominant uncertainty.
- Knowing the flavour of each jet will reduce the *flavour composition* component of this uncertainty.



# Why do we want a hadron-level jet classifier?

---

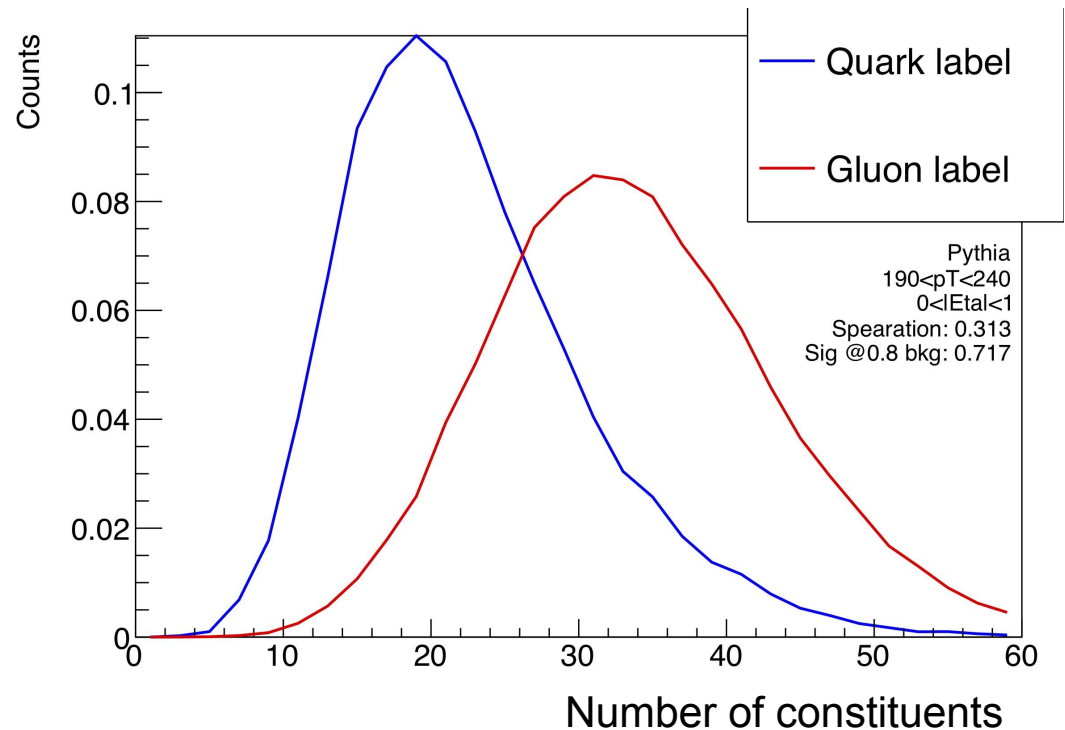
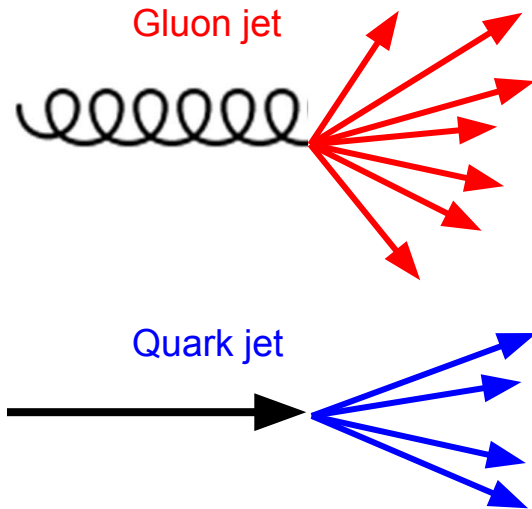
- Jets are currently tagged as quark or gluon jets by looking at the highest  $p_T$  parton at parton shower level.
- Parton showering is modelled by a Monte Carlo generator such as Pythia, Herwig, Sherpa.
- Partons are unphysical, so the parton level is handled differently by each generator.
- Using partons to label jets introduces a dependence on the generator used. Instead we should define quark and gluon jets by their ***physical properties at hadron level***. This is the focus of my work.

# How can we distinguish quark/gluon hadron jets?

---

# How can we distinguish quark/gluon hadron jets?

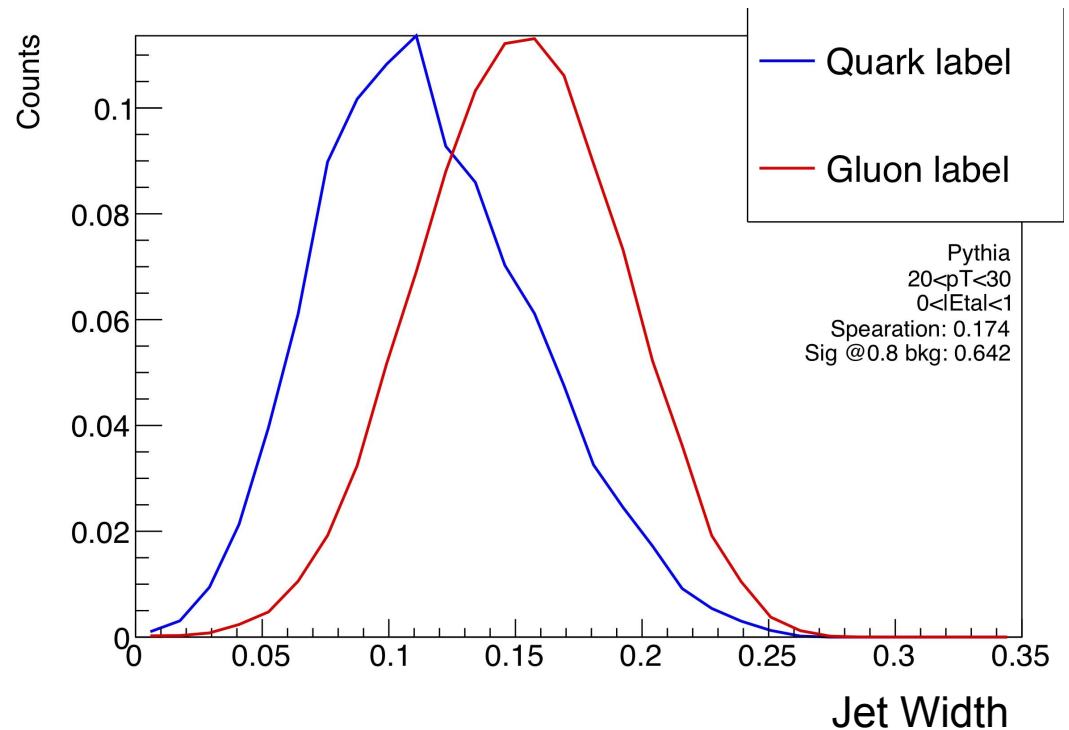
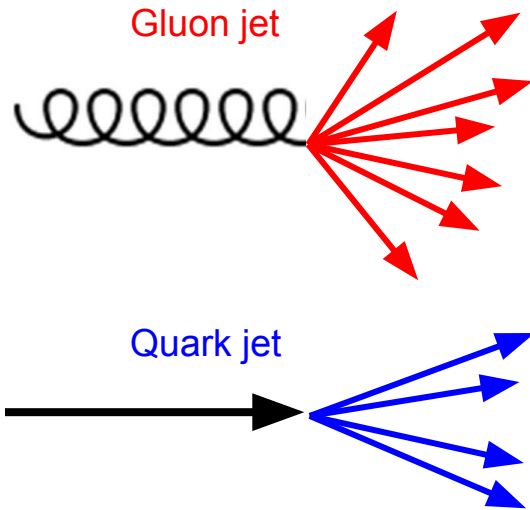
- Number of constituents:  
On average, gluon jets have more constituent particles.



# How can we distinguish quark/gluon hadron jets?

- Number of constituents.
- Width:  
On average, gluon jets are wider.

$$w = \frac{\sum_i p_{T,i} \times \Delta R(i, \text{jet})}{\sum_i p_{T,i}}$$



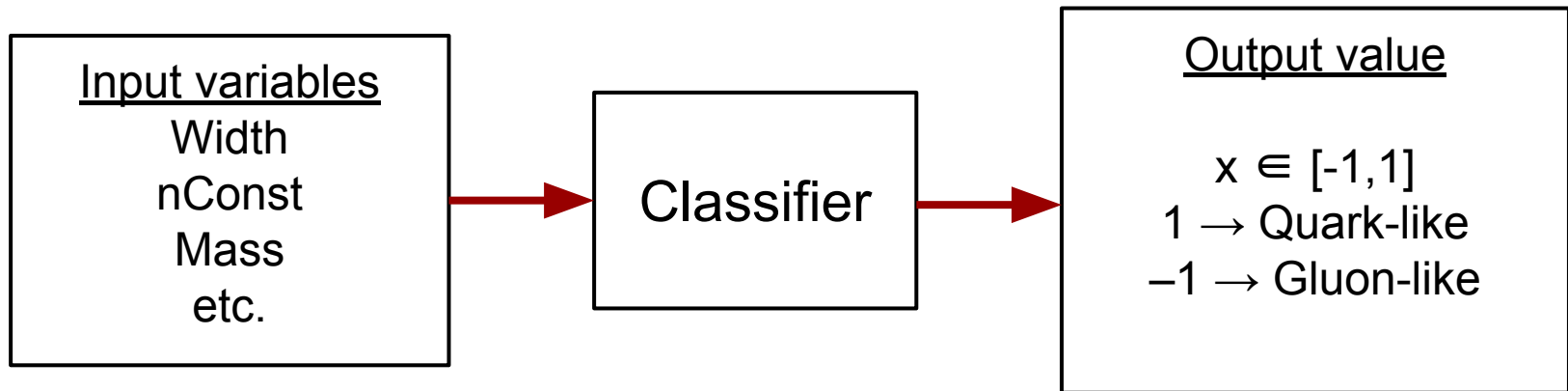
# How can we distinguish quark/gluon hadron jets?

---

- Number of constituents.
- Width.
- ...

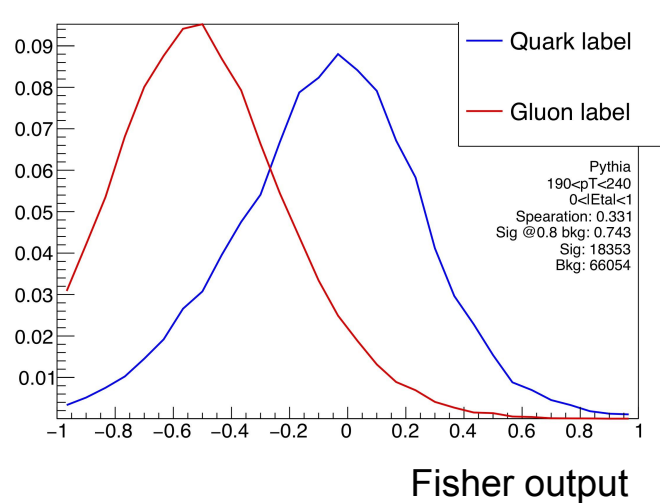
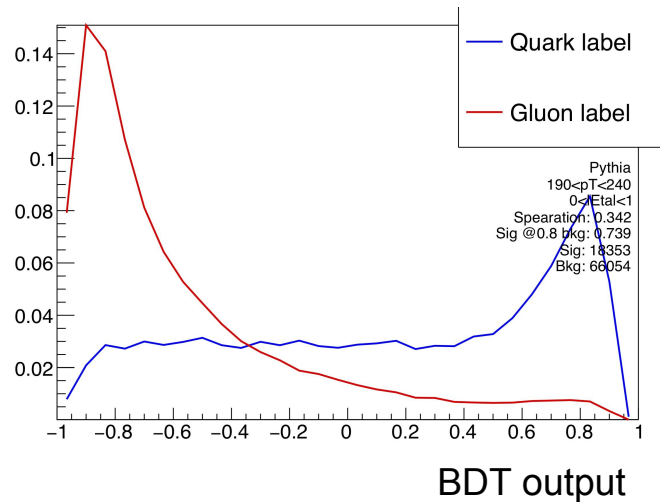
Total of 13 input variables

# Training a classifier using TMVA



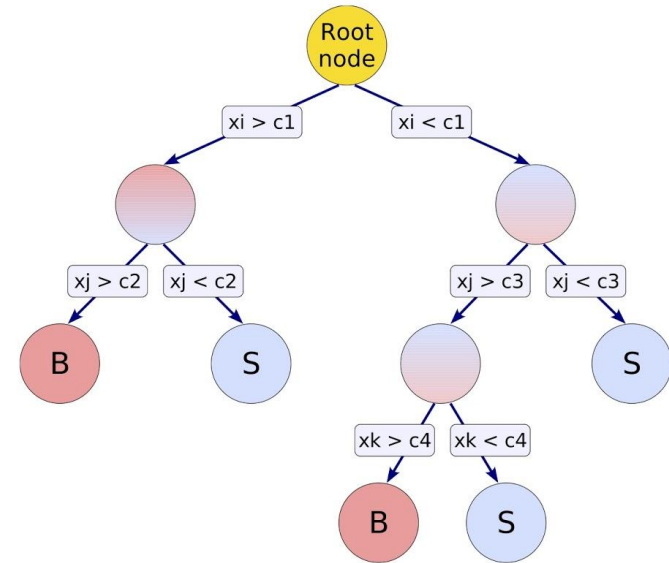
- Classifier trained to distinguish two classes of data: quark and gluon jets (as labelled using parton information).
- Gives each jet a value on  $[-1,1]$  depending if it is “quark-like” or “gluon-like”.

# Two MVA methods used



← Gluon-like                      Quark-like →

## Boosted Decision Trees



## Fisher

- Linear discriminant.
- Training data used to determine coefficients.

$$y_F(i) = F_0 + \sum_{k=1}^{n_{\text{var}}} F_k x_k(i)$$

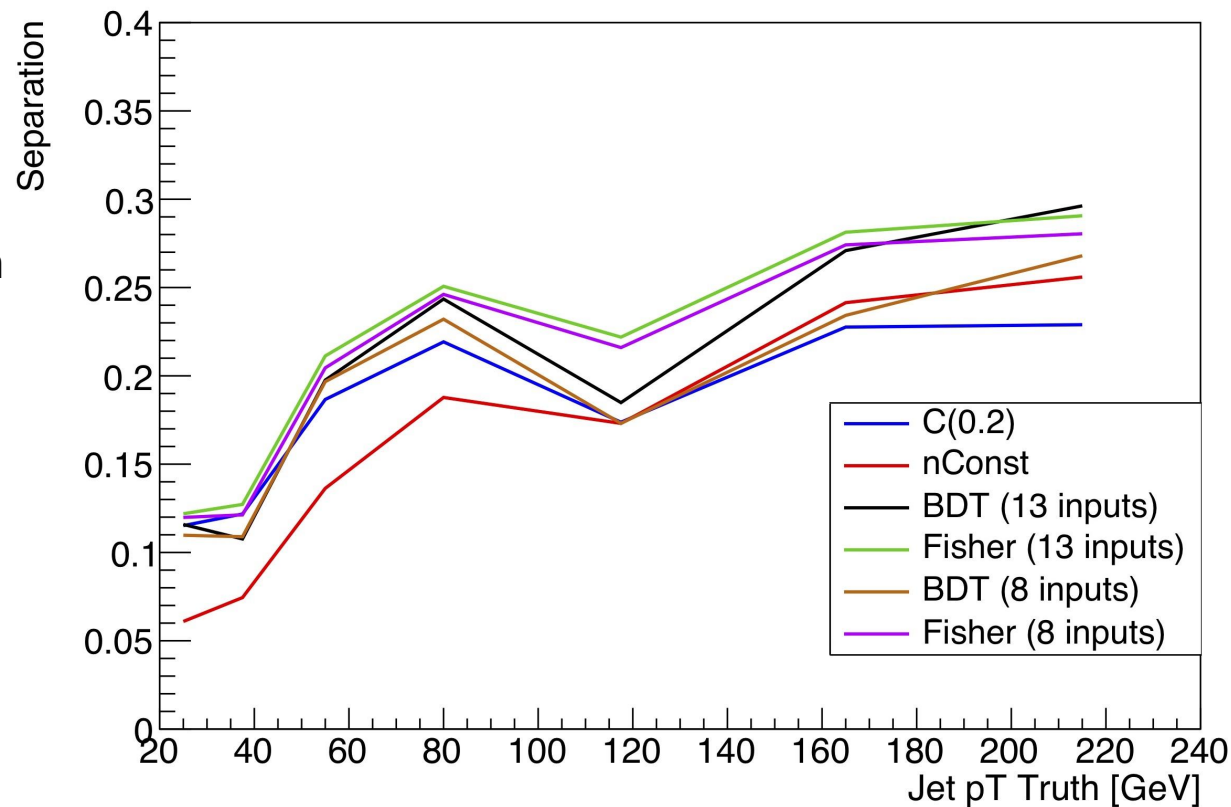


# How effective are the classifiers?

A good metric is the separation between quark and gluon distributions.

$$\langle S^2 \rangle = \frac{1}{2} \int \frac{(\hat{y}_S(y) - \hat{y}_B(y))^2}{\hat{y}_S(y) + \hat{y}_B(y)} dy$$

- Is 1 for two distributions with no overlap, and 0 for two identical distributions.
- Trained classifiers are stronger than the strongest input variable.
- Fisher is as good as or better than Boosted Decision Trees.



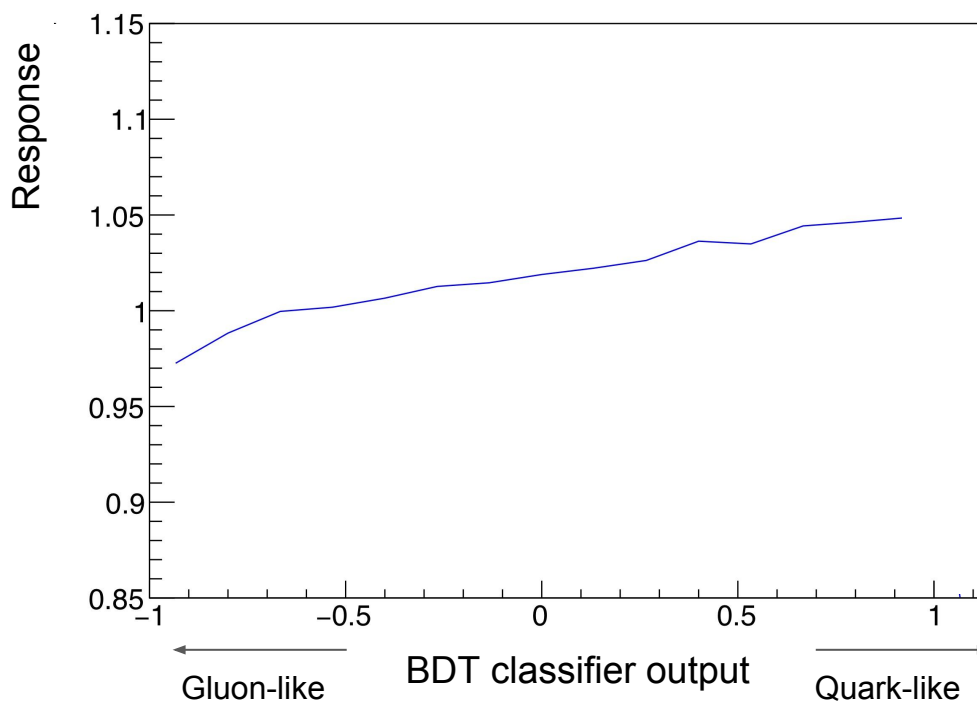
# Response of “quark-like” and “gluon-like” jets

---

- Quark jets have a higher average response than gluon jets.
- We would hope that the more “quark-like” a jet is, as defined by our classifier, the higher the response.

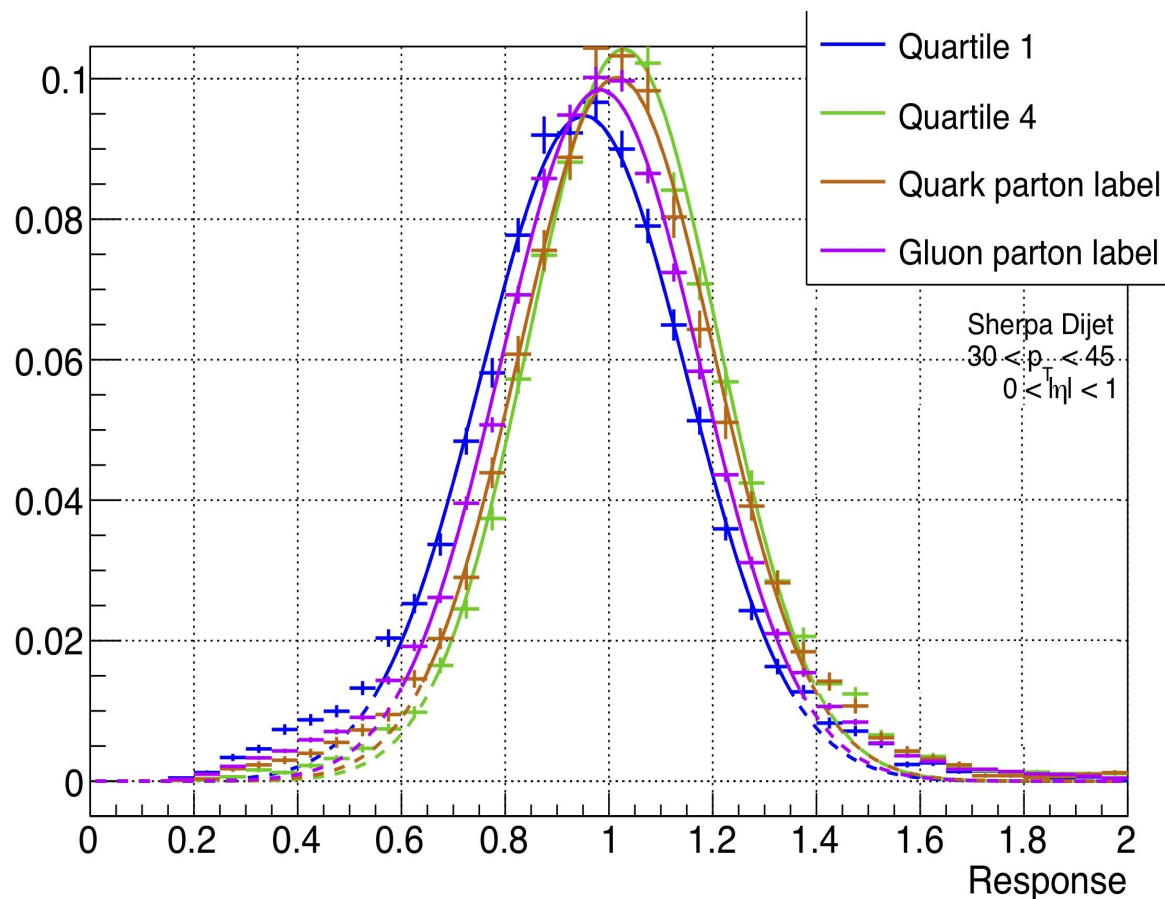
# Response of “quark-like” and “gluon-like” jets

- Quark jets have a higher average response than gluon jets.
- We would hope that the more “quark-like” a jet is, as defined by our classifier, the higher the response.



It works  
very well!

# Response of “quark-like” and “gluon-like” jets



Gluon jets labelled using parton information.

Quark jets labelled using parton information.

25% most “gluon-like” jets using classifier.

25% most “quark-like” jets using classifier.

Classifier can describe the response difference between quark and gluon jets just as well as the parton information can!

# Conclusions

---

- I have trained quark/gluon jet classifiers using hadron level inputs.
  - Obtain ~70% quark-jet efficiency at 80% gluon-jet rejection
- A simple Fisher discriminant is as effective as using Boosted Decision Trees.
- The “quark-iness” or “gluon-iness” of a jet is linearly related to its calorimeter response.
  - Can hence be used to parameterize JES uncertainty!

# My CERN experience



- IPP/CERN summer student program.
- ~300 summer students from ~100 countries.
- 5 week lecture series.
- Experiment tours.
- Europe travel.



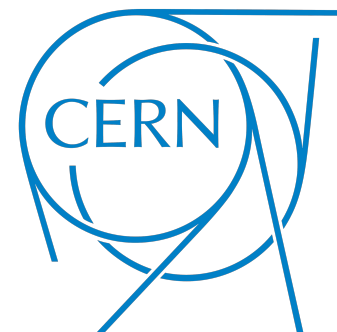
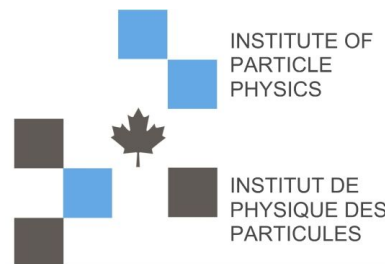
- Presenting to and getting feedback from other groups.
- ROOT, machine learning, data.
- Working independently on a research project.
- Time management between work and fun.

Many thanks to:

My supervisor Dag Gillberg!

Ben Nachman and the q/g tagging group

The JES/JER group



# Future work

---

- Create a software tool to decorate jets with a hadron level quark/gluon label.
- Use this tool as part of ATLAS jet reconstruction so all jets are labelled.
- Use this label in JES uncertainty parameterization.



# How can we distinguish quark/gluon hadron jets?

- Number of constituents.
- Width.
- Mass.
- Fraction of jet  $p_T$  carried by:
  - Charged hadrons
  - Photons
  - The highest- $p_T$  hadron
- Total jet charge, weighted by  $p_T^{0.5}$ .
- Number of constituents carrying 90% of jet  $p_T$ .

- Effective number of constituents:

$$N_{\text{const}}^{\text{eff}} = \frac{(\sum_i p_{T,i})^2}{\sum_i p_{T,i}^2}$$

- Jet energy sharing value:

$$p_T D = \frac{\sqrt{\sum_i p_{T,i}^2}}{\sum_i p_{T,i}}$$

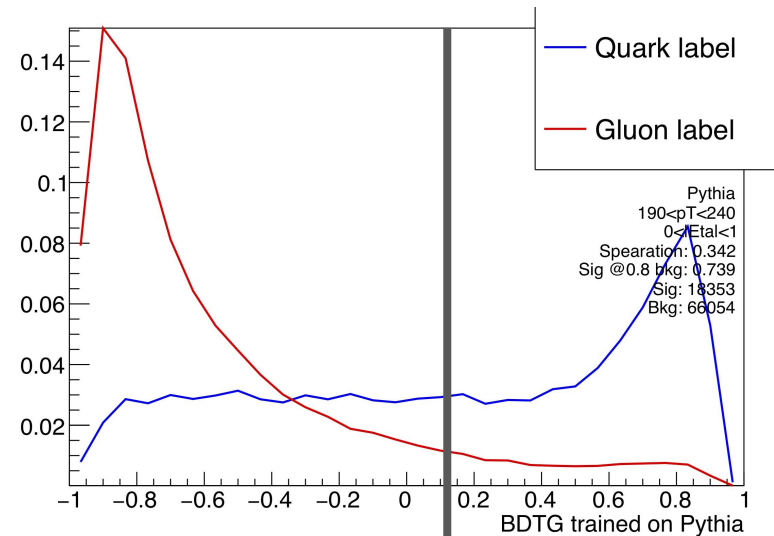
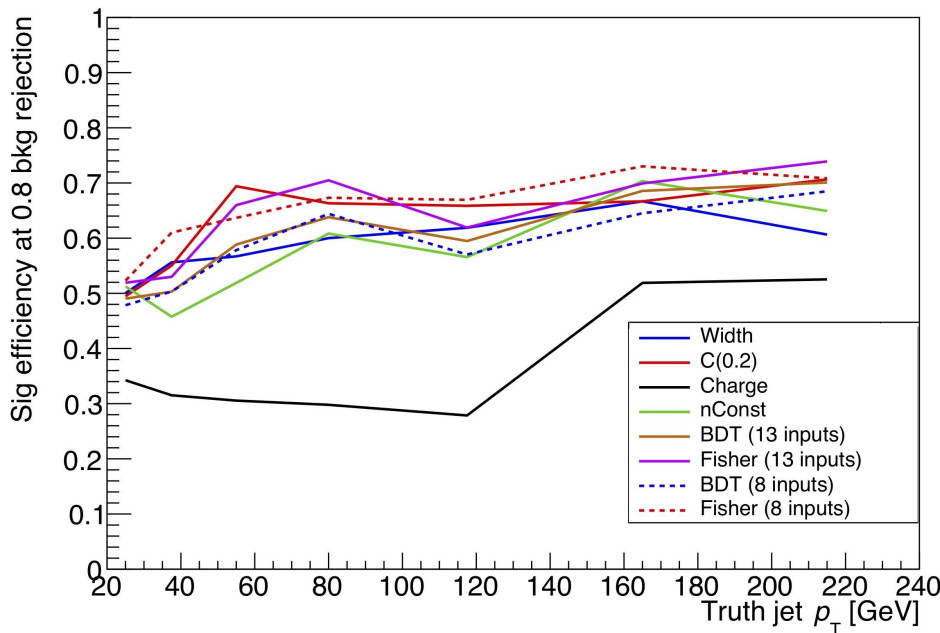
- Energy-energy-correlation angularity with  $\beta = 0.2, 1.0, 2.0$ .

$$C(\beta) = \frac{\sum_i \sum_j p_{T,i} \times p_{T,j} \times (\Delta R(i, j))^\beta}{(\sum_i p_{T,i})^2}$$

# How effective are the classifiers?

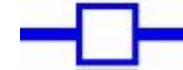
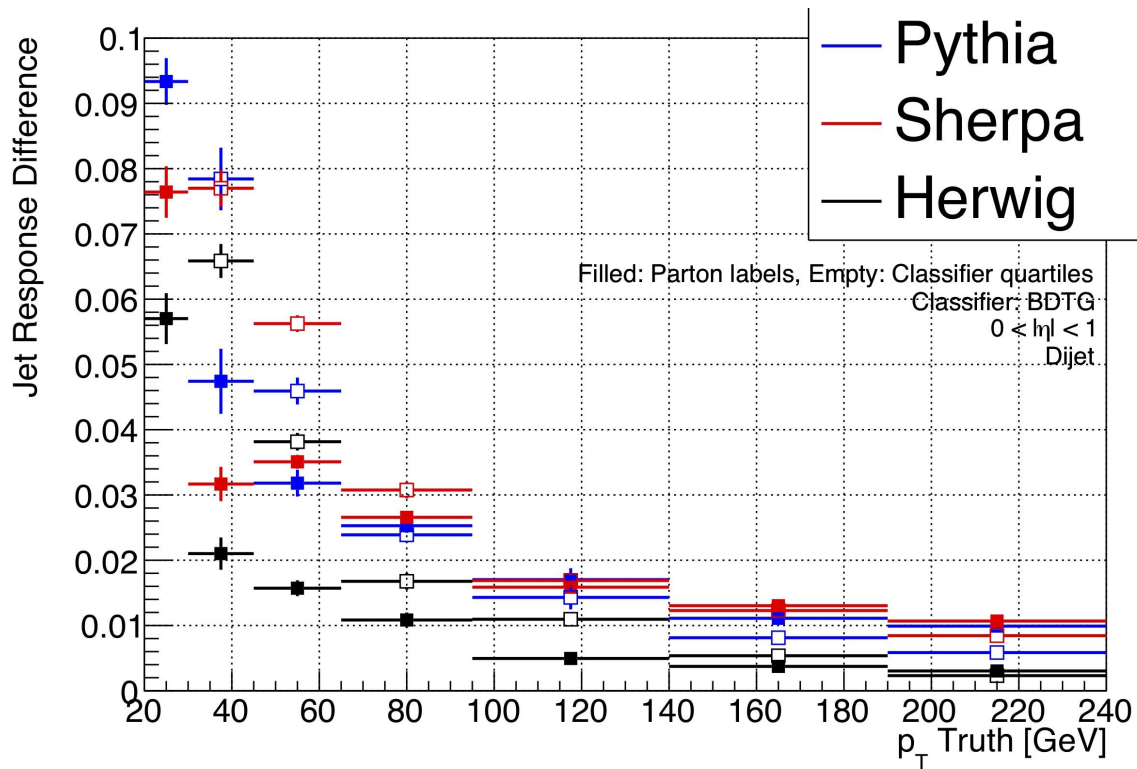
One metric is the quark-jet efficiency at 80% gluon-jet rejection.

1. Make a cut so that 80% of the gluon jets are excluded.
2. What fraction of quark jets survive?



80% gluon-jets rejected  
60% quark-jets remaining

# Have we avoided generator dependance?

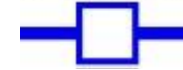
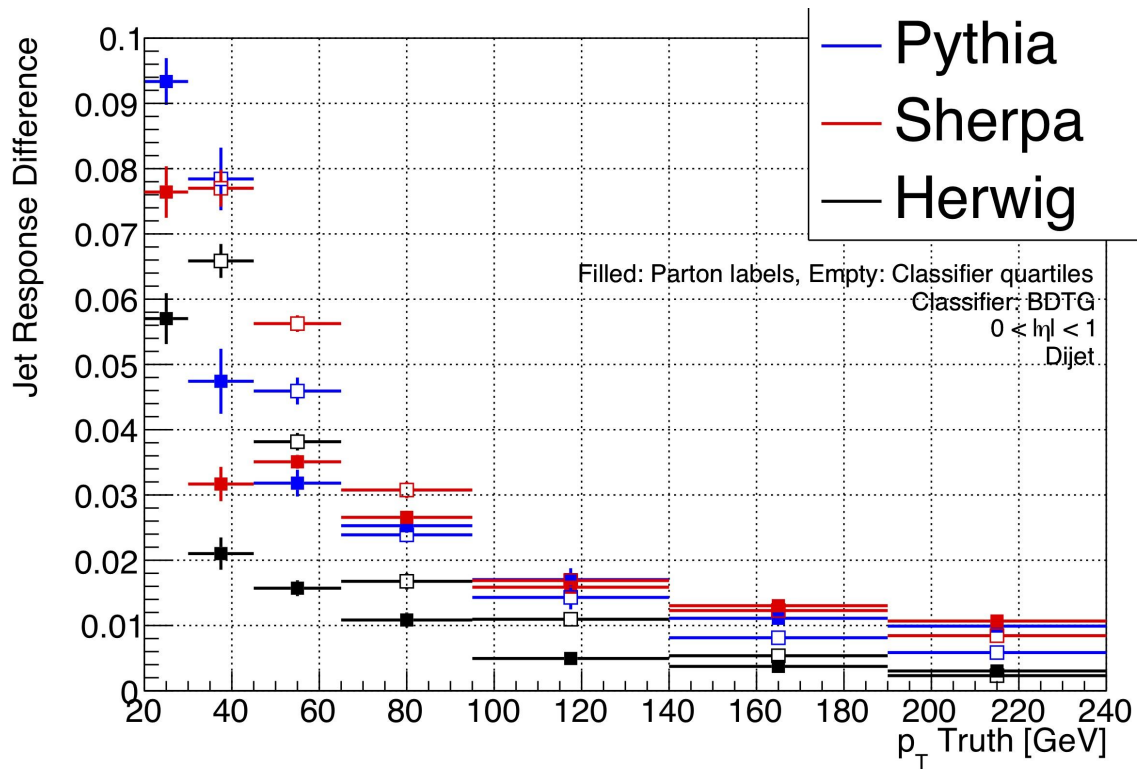


Distance between classifier peaks on previous slide



Distance between parton label peaks on previous slide

# Have we avoided generator dependence?



Distance between classifier peaks on previous slide



Distance between parton label peaks on previous slide

Not yet



- Preliminary results: still a spread between empty bubbles of different colours. More work is needed.
- Generator-dependent parton information was still used for training – might need a new approach to see improvement.