

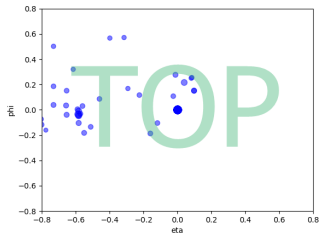
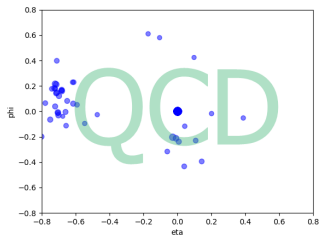
How to find new physics without physicists

Simon Kluettermann

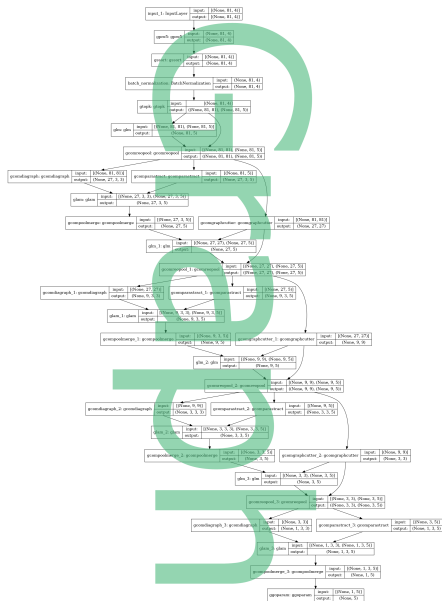
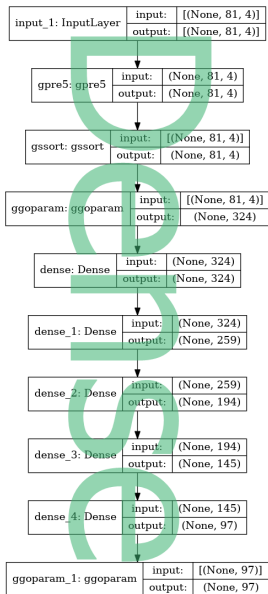
12. Februar 2021

Top Tagging

- Looking at LHC jets (particle showers)
- Millions of jets, maybe some of them weird in some way
- But at most a few and in some unknown way
- So use ML to filter them out
- Usually tested by trying to find top jets in a qcd background



- ParticleNet
 - supervised
 - Graph Neuronal Networks
- Qcd Or What?
 - unsupervised
 - Convolutional Networks
- Combine ideas from both
- Into a Graph Autoencoder



Assuming

The higher the loss of the comparison network

The more likely this is a top jet.

You reach an AUC of

0.908

- Already fairly satisfied
- If I beat this, my graph networks work
- AUC score
- between 0 and 1, higher=better
- 1:perfect,0.5:random

Assuming

The higher the loss of my graph network

The more likely this is a top jet.

You reach an AUC of

0.910

- So my network works
- But...
- AUC score
- between 0 and 1,
higher=better
- 1:perfect,0.5:random

Assuming

The higher R^2

The more likely this is a top jet.

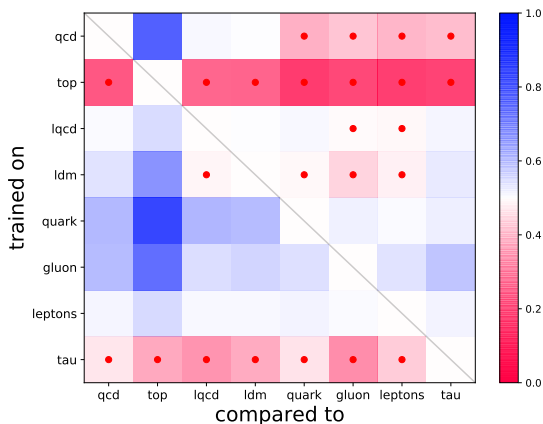
You reach an AUC of

0.915

- trivial network
- best score yet
- AUC score
- between 0 and 1,
higher=better
- 1:perfect,0.5:random

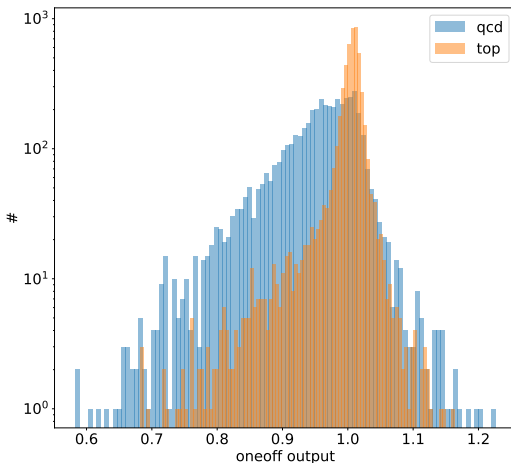
Results

- The more blue the better, and if a pixel is red (dot) it is not detectable
- For the comparison work only 10(32)/56 are detectable
- useless, except for qcd vs top



OneOff Networks

- So improve it using my own algorithm:
- Train a network to output a constant
- $loss = (f(x) - 1)^2$
- Anomalous data usually does not reproduce the same constant



Results

- The more blue the better, and if a pixel is red (dot) it is not detectable
- Quality is not final, but
- here do 48(52)/56 comparisons work

