

Sydney group update

Australian ATLAS Meeting: August 19, 2021



THE UNIVERSITY OF
SYDNEY

Introduction: Four Tops Analysis

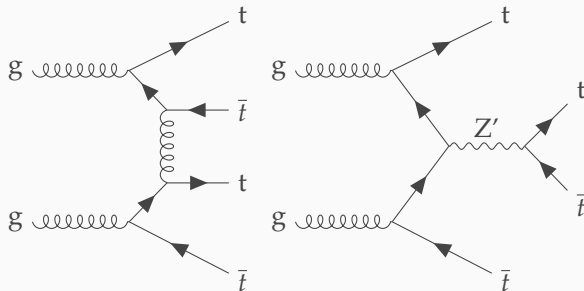


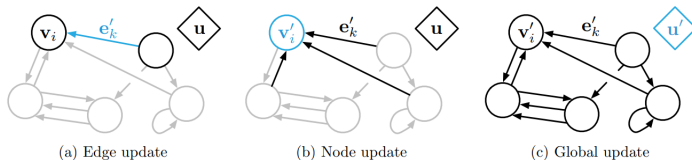
Figure 1: $t\bar{t}t\bar{t}$ production Feynman diagrams for left) Standard Model right) BSM Top-Philic model

- Within the SM, the 4-top production is extremely rare ($\sigma = 12.6^{+5.6}_{-5.2}\text{fb}$) and occurs via gluon fusion.
- Some BSM physics models predict a boosting of this cross-section from modified top quark couplings or heavy resonances (Figure 1).

Introduction: Four Tops Analysis

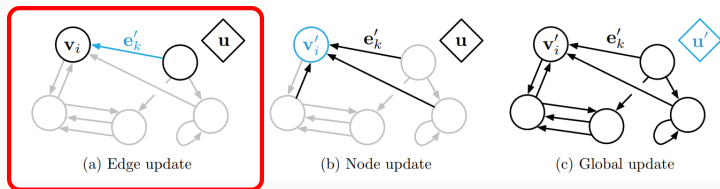
- Products of resonance tops are expected to be largely boosted and produce fat-jets.
- These jet types can be reclustered into so called RC-jets in single lepton analyses.
- Graph Neural Networks (GNNs) could potentially assist in identifying jets originating from resonance tops.
- Although the data science technique has been around since 2009, it only recently gained mainstream and HEPP attention.
- A similar problem was partially solved using GNNs to group jets to common parents in the $t\bar{t}H$ production at Berkeley.

What is a Graph Neural Network?



- A Graph Neural Network is a generalized deep learning technique similar to a conventional Neural Network (Deep Layers, Aggregation, Convolution, Pooling etc.)
- What makes a GNN unique is that, non-euclidean sampled data can be encoded on graph data structures where in the context of HEPP;
 - Nodes - Some object (Particle, Jets) that has attributes (η, ϕ, p_T, \dots)
 - Edges - Some relation between objects; ΔR , Inv Mass, Δ^* .
 - Graph - Some global properties of the graph; Collision Energy, Missing p_T , $\phi \dots$

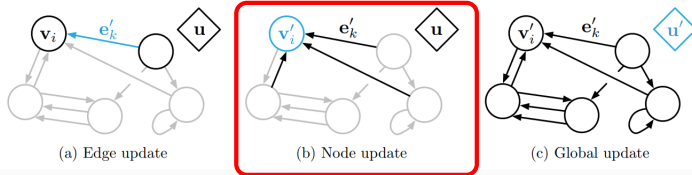
What is a Graph Neural Network?



Given some graph data structure, with edge and node features, an example computation would look something like this:

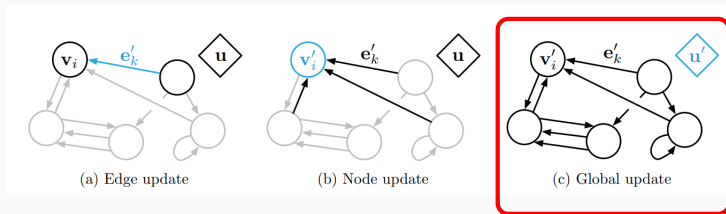
- Trainable edge weights are first updated via a multilayer-perceptron (MLP).
- The inputs of the MLP are the current edge state, attributes from *sender* and *receiver* nodes, and some global value.
- Edges are finally updated through a permutation invariant function; summation, averaging, max, min, ...
- This allows for edges to be mapped appropriately to nodes.

What is a Graph Neural Network?



- Nodes are updated by applying an MLP to the node features and updated incoming edge attributes.
- Using the updated edges and nodes of the graph, a permutation invariant function is again used to map to a global value or attribute.

What is a Graph Neural Network?



- Finally, a global attribute update is performed with the global attribute value and updated edge and node vectors.
- The described sequence is **not** necessarily enforced, but can be adjusted arbitrarily, adding to the flexibility of GNNs.
- This flexibility has opened many new architectures; Auto-encoding, Node/Edge predictions, Graph Convolution Neural Networks, Multi-Block models, ...

My Current Progress in Implementing GNNs

- I have been writing most of the infrastructure code;
 - Reading multiple ROOT files (containing resonance events)
 - Converting particle objects; jets, electrons, muons, truth particles, tops, ... into python objects for matching
 - Matching truth particles to truth jets and matching truth jets to detector jets using a ΔR matching technique.
 - Integrated a very basic working GNN into the code using pytorch-geometric.
- But there are still some issues that need resolving:
 - Truth jet and truth particle matching using ΔR is not very unreliable (likely due to gluons and any quarks below charm given a flavor value of 0).
 - Jet and truth jet matching is also experiencing the same issue.
- Next things to work on:
 - Exploring more GNN architectures and comparing their performance on the truth children (perfectly matching tops).
 - Understanding how to properly match jets and truth jets to truth children.

My Current Progress in Implementing GNNs

```
>>>>===== New Event =====>>>>===== DetectorMatchingEngine =====>>>>
--- All Particles ---
Type: jet FL: 5 -- FLE: 55 --- TruthParton: 5
Type: jet FL: 15 -- FLE: 15 --- TruthParton: 1
Type: jet FL: 0 -- FLE: 0 --- TruthParton: 21
Type: jet FL: 0 -- FLE: 0 --- TruthParton: 21
Type: jet FL: 0 -- FLE: 0 --- TruthParton: 2
Type: jet FL: 5 -- FLE: 5 --- TruthParton: 5
Type: jet FL: 0 -- FLE: 0 --- TruthParton: 2
Type: jet FL: 4 -- FLE: 4 --- TruthParton: 4
Type: jet FL: 5 -- FLE: 5 --- TruthParton: 5
Type: jet FL: 0 -- FLE: 0 --- TruthParton: 21
Type: jet FL: 0 -- FLE: 0 --- TruthParton: 21
Type: nu TruTyp: 6 TruOri: 10 Prompt: 1 C: -1.0
Type: truthjet FL: 5 FLE: 55 nChad: 1 nBHad: 2 Index: 0
Type: truthjet FL: 0 FLE: 0 nChad: 0 nBHad: 0 Index: 1
Type: truthjet FL: 0 FLE: 0 nChad: 1 nBHad: 0 Index: 2
Type: truthjet FL: 0 FLE: 0 nChad: 0 nBHad: 0 Index: 3
Type: truthjet FL: 5 FLE: 5 nChad: 1 nBHad: 1 Index: 4
Type: truthjet FL: 0 FLE: 0 nChad: 0 nBHad: 0 Index: 5
Type: truthjet FL: 0 FLE: 0 nChad: 0 nBHad: 0 Index: 6
Type: truthjet FL: 4 FLE: 4 nChad: 1 nBHad: 0 Index: 7
Type: truthjet FL: 5 FLE: 5 nChad: 0 nBHad: 1 Index: 8
Type: truthjet FL: 15 FLE: 15 nChad: 0 nBHad: 0 Index: 9
Type: truthjet FL: 0 FLE: 0 nChad: 0 nBHad: 0 Index: 10
Type: truthjet FL: 0 FLE: 0 nChad: 0 nBHad: 0 Index: 11

##### dR #####
TJet Flavour: 0 Extended: 0 nChad 0 nBhad 0 Index: 3 Jet Flavour: 0 Extended: 0 Truth Parton: 2 HS: 1 dR: 0.0096
TJet Flavour: 5 Extended: 55 nChad 1 nBhad 2 Index: 0 Jet Flavour: 5 Extended: 55 Truth Parton: 5 HS: 1 dR: 0.0108
TJet Flavour: 0 Extended: 0 nChad 1 nBhad 0 Index: 2 Jet Flavour: 0 Extended: 0 Truth Parton: 21 HS: 1 dR: 0.013
TJet Flavour: 5 Extended: 5 nChad 1 nBhad 1 Index: 4 Jet Flavour: 5 Extended: 5 Truth Parton: 5 HS: 1 dR: 0.0158
TJet Flavour: 5 Extended: 5 nChad 0 nBhad 1 Index: 8 Jet Flavour: 5 Extended: 5 Truth Parton: 5 HS: 1 dR: 0.0167
TJet Flavour: 0 Extended: 0 nChad 0 nBhad 0 Index: 19 Jet Flavour: 0 Extended: 0 Truth Parton: 21 HS: 1 dR: 0.0208
TJet Flavour: 0 Extended: 0 nChad 0 nBhad 0 Index: 1 Jet Flavour: 0 Extended: 0 Truth Parton: 21 HS: 1 dR: 0.0222
TJet Flavour: 0 Extended: 0 nChad 0 nBhad 0 Index: 5 Jet Flavour: 0 Extended: 0 Truth Parton: 2 HS: 1 dR: 0.0235
TJet Flavour: 4 Extended: 4 nChad 1 nBhad 0 Index: 7 Jet Flavour: 4 Extended: 4 Truth Parton: 4 HS: 1 dR: 0.0447
TJet Flavour: 0 Extended: 0 nChad 0 nBhad 0 Index: 11 Jet Flavour: 0 Extended: 0 Truth Parton: 21 HS: 1 dR: 0.0772
TJet Flavour: 0 Extended: 0 nChad 0 nBhad 0 Index: 6 Jet Flavour: 15 Extended: 15 Truth Parton: 1 HS: 1 dR: 0.118
TJet Flavour: 15 Extended: 15 nChad 0 nBhad 0 Index: 9 Jet Flavour: 15 Extended: 15 Truth Parton: 1 HS: 1 dR: 0.2895
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TJet Flavour: 0 Extended: 0 nChad 0 nBhad 0 Index: 3 Jet Flavour: 15 Extended: 15 Truth Parton: 1 HS: 1 dR: 4.9736
```

Thanks for Listening!

Peter W. Battaglia et al. “Relational inductive biases, deep learning, and graph networks”. In: *CoRR* abs/1806.01261 (2018). arXiv: 1806.01261. URL:

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<http://dx.doi.org/10.1140/epjc/s10052-019-7593-7> Jeong Han Kim et al. “Probing TeV scale top-philic resonances with boosted top-tagging at the high luminosity LHC”. In: *Physical Review D* 94.3 (2016). ISSN: 2470-0029. DOI: 10.1103/physrevd.94.035023. URL: <http://dx.doi.org/10.1103/PhysRevD.94.035023>