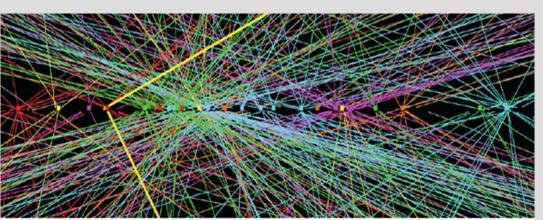


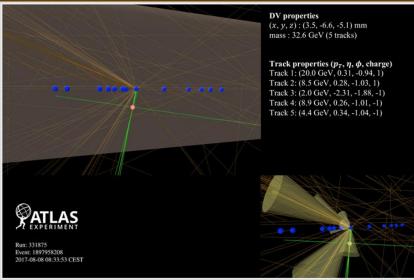
TrackOpt vertex project update

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Reminder







Problem:

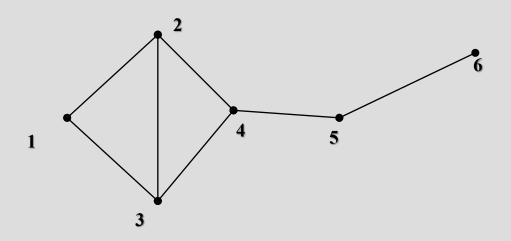
Having a set of tracks find all vertices (track production points)

Solution idea:

- 1. Create a track-track compatibility graph using a priory physics knowledge, i.e. explicitly calculate a point of closest approach of the 2 tracks in 3D space and the track/vertex parameters at this point.
- 2. Based on this information, estimate a probability that a given 2-track vertex is real
- 3. Use LMC algorithm to partition this graph with weighted edges

Graph





Example: 6 tracks (nodes)

7 possible 2-track vertices (edges)

Node vector of features(track parameters): td0, tZ, tPhi, tEta, tQoP, tTheta, tChi2, tNDoF, tCovD0, tCovZ,tCovD0Z, tSignif Edge vector of features(2-track vertex): vChi2, vX, vY, vZ, vcovXX, vcovXY, vcovYY, vcovXZ, vcovYZ, vcovZZ, vsumPt, vEta, vPhi, vDZ, massPiPi, vtrue, vCharge, isGamma, isKs, isLambda + truth_label

Prepared graphs saved in ROOT format (~1.5mb/ev, ~1400nodes/ev, ~26k edges/ev)

For the moment, to save CPU, for tracks closer that 3σ to the beamline no real vertex fit is done. Compatibility is calculated based on Z track position on the beamline (1D fit, like in the PV finding paper).

Edge weights with DGL



Edge weights estimation is implemented in DGL

- 1) Node_hidden_state = NN{Concat(Node_features, Mean(Edge_features))}
- 2) Edge_weight = NN{ Concat(Node_hidden_state_i, Edge_features, Node_hidden_state_j)}



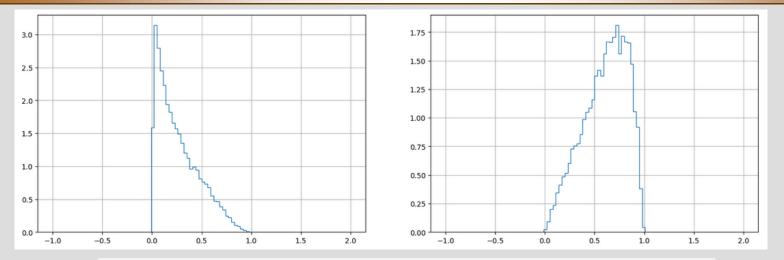


- 3) Node hidden state = NN{Concat(Node features, Weighted Mean(Edge features)}
- 4) Upd_Edge_weight = NN{ Concat(Node_hidden_state_i, Edge_features, Node_hidden_state_j)}

Steps (3),(4) happened to be not needed, final weights are practically the same.

Loss – binary cross-entropy (classification) Activations – Mish + Sigmoid to get probability

DGL results (very preliminary)



AUC ~ 0.84 for both primary and secondary edges(aka. 2-track vertices)

For comparison - XGBoost classification of edges based on the same edge features: AUC ~ 0.75

Next steps

- 1) Save training results and run LMC on weighted graphs
- 2) Implement metrics for comparison
- 3) Play with DGL
 - ✓ More graphs for training (currently batch of 50)
 - ✓ Different layers LSTM, Transformers, etc,
 - ✓ Different reduction functions
 - ✓
- 4) More into the future try edge prediction approach based on node(track) features only. This will allow to estimate a performance gain due to a priory knowledge injected as edge features
 - Similarity to foundation model