Particle Track Analysis using Neural Networks



Quick update

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NN on full unclustered events

| | | Clus | Cluster 2 | | | | | | | Cluster 3 | | | | | | | | | |
|----------------|-------|-------|-----------|-------|-----------|-------|-------|-------|-------|-----------|--|-------|-------|-------|-------|-------|-------|-------|-------|
| Strip pos (mm) | 728,2 | 728,6 | 729,0 | 729,3 | 741,2 | 741,6 | 742,0 | 742,4 | 742,7 | 743,1 | | 745,3 | 745,7 | 746,1 | 746,4 | 746,8 | 747,2 | 747,6 | 747,9 |
| Charge | 203 | 260 | 102 | 116 | 252 | 131 | 76 | 92 | 83 | 91 | | 110 | 187 | 274 | 388 | 274 | 283 | 103 | 109 |
| Timing (ns) | 94 | 86 | 54 | 37 | 25 | 30 | 39 | 43 | 71 | 67 | | 17 | -5 | 27 | 23 | 47 | 52 | 66 | 93 |

Example of an event with multiple clusters

NN trained with full unclustered events of a testbeam dataset with 29 degrees inclination « True » position of the particle is ~746.8mm (cluster 3).

Charge weighted position on the whole event : 741.6mm, charge weighted position on cluster 3 : 746.6mm. NN manages to reconstruct the true position with a residual of 50µm à priori ignoring cluster 1 and 2.



Transformer compared to MLP

Testbeam 29 degrees 530V 100ns



Results are minimally better. Will try transformers with more parameters. In general we seem to hit a wall at \sim 130µm core resolution for 29 degree inclination.

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Backup – Transformer architecture



Example of a transformer regression architecture (source)

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Attention mechanism (paper) Model assign weights to compare how each input feature stands to each other.