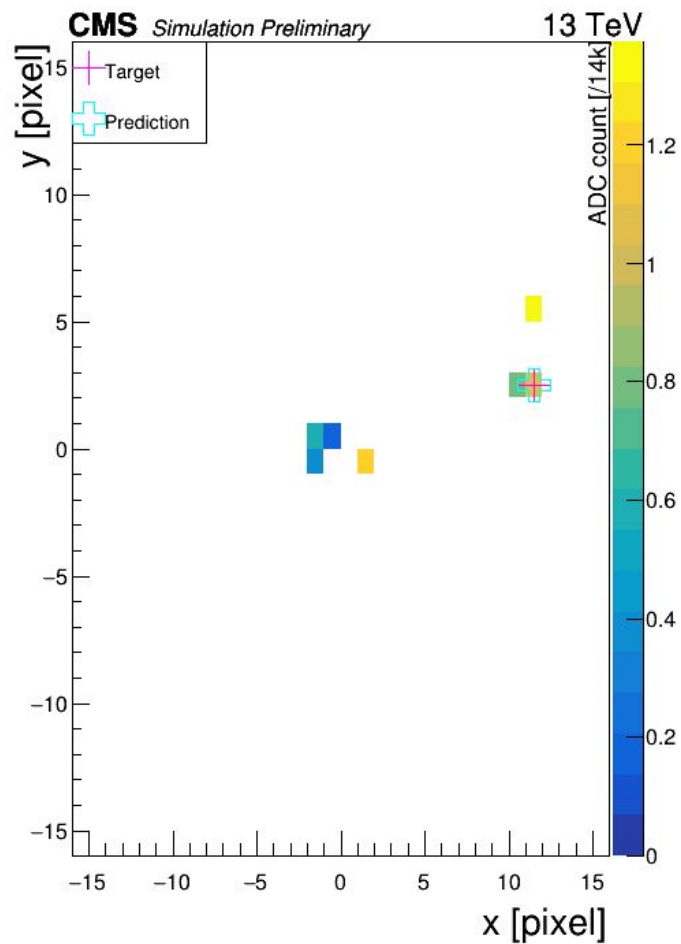


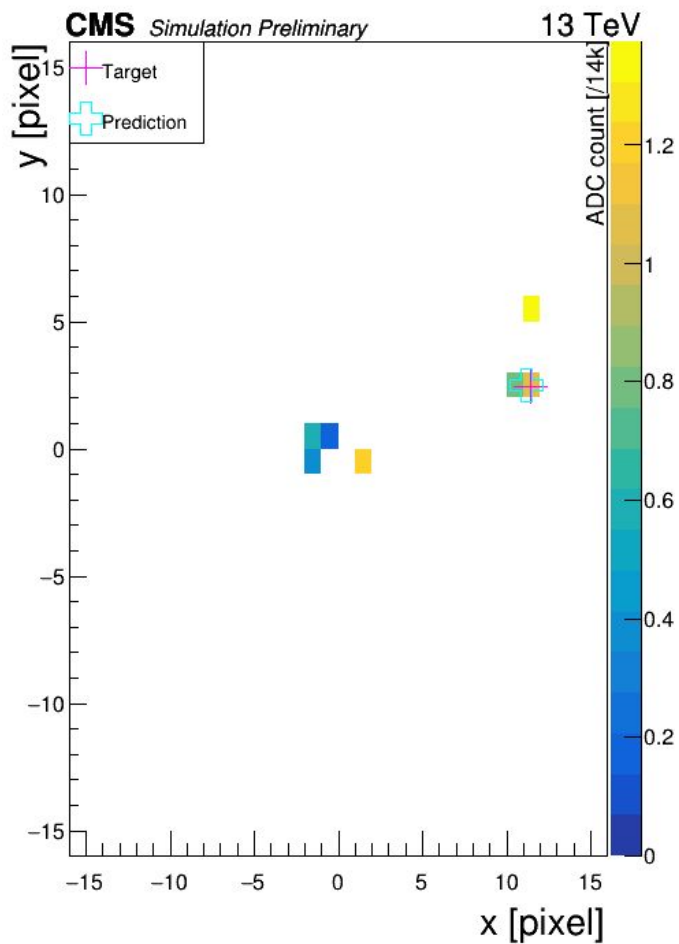
Résumé of Summer Research

Liam O'Shaughnessy - 25/09/23

Pixel Window, layer 2



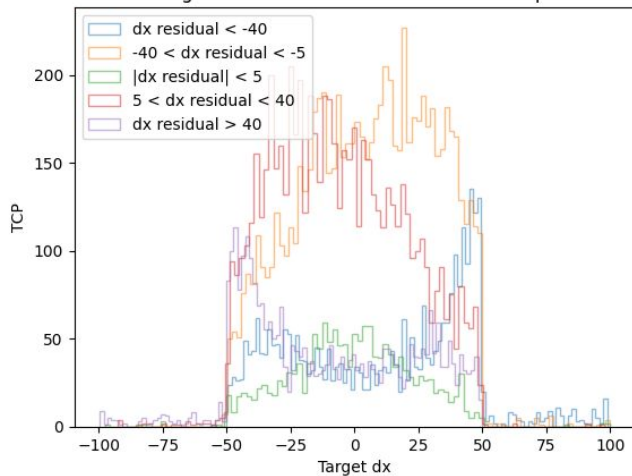
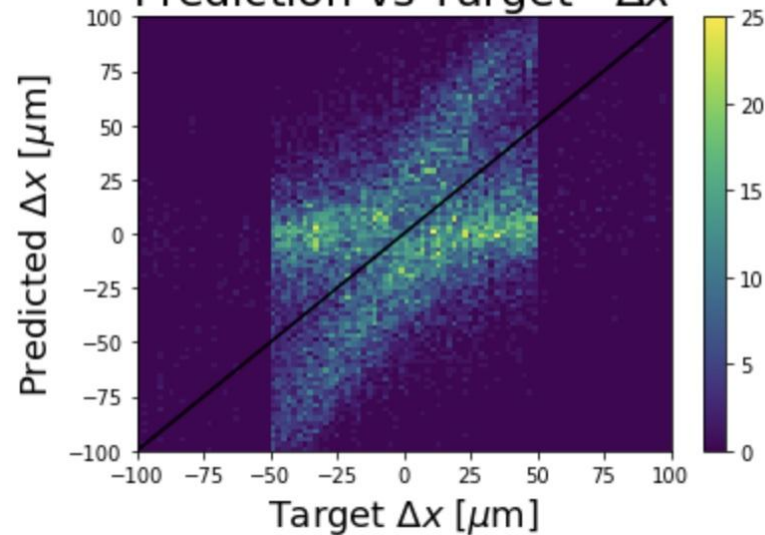
Pixel Window, layer 2



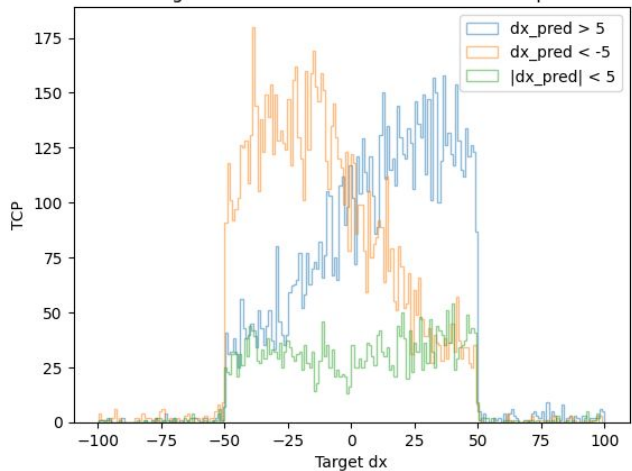
Summary

- Have been examining target values of track crossing point parameters (dx, dy, jet eta, etc.) compared to DeepCore predictions of these parameters
- Have been looking at interesting cases where DeepCore fails to predict correctly and why e.g. the dx boomerang, the TCP plots with odd boundaries
- Trying to break these down to see if odd splits can be explained by TCP's being in certain parameter regime, having charge issues, etc.

Target dx Distribution - Residual dx Groups

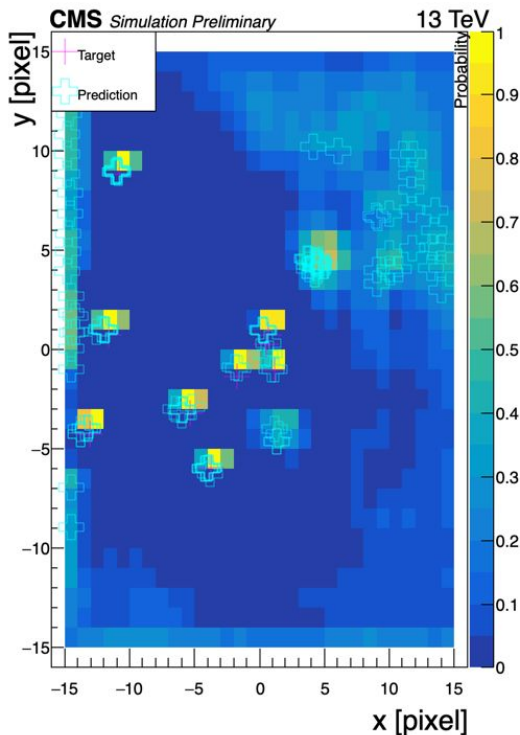
Prediction vs Target - Δx 

Target dx Distribution - dx Prediction Groups

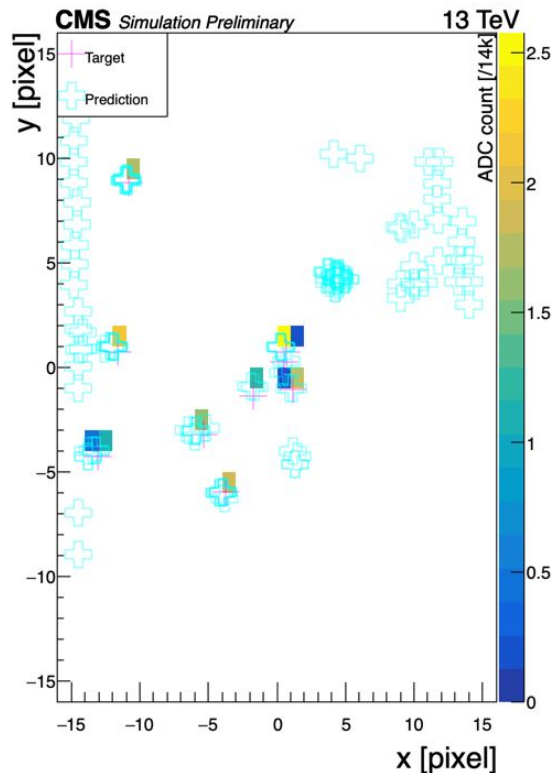


Tried to split the boomerangs by residual, by positive/negative/zero dx prediction, by eta and pt -> does not appear to cleanly split.

TCP Prediction Map, overlap 0

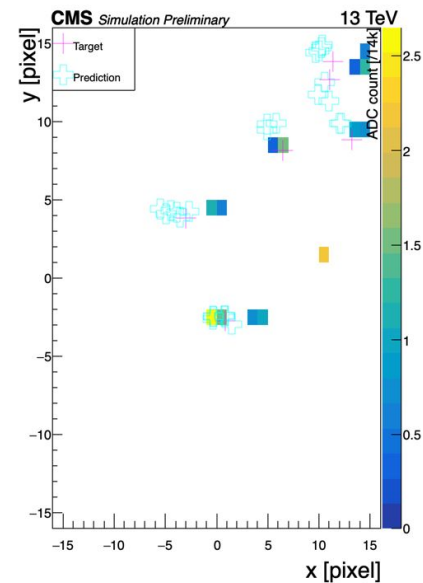


Pixel Window, layer 2

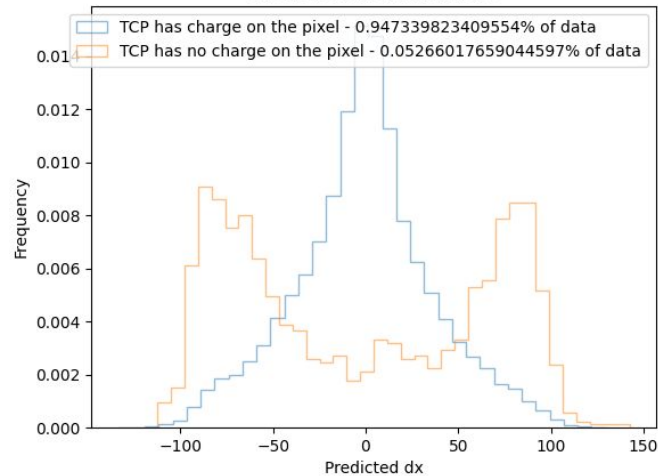


From the DeepCore visualizer, these had a lot of bad predictions at lower thresholds along the edges, tried to figure out what was going on here. When I raised the threshold on the TCP check, I believe many of the weird ones disappeared, but can recheck this. This also prompted question about if the linear propagation is causing issues/is way off.

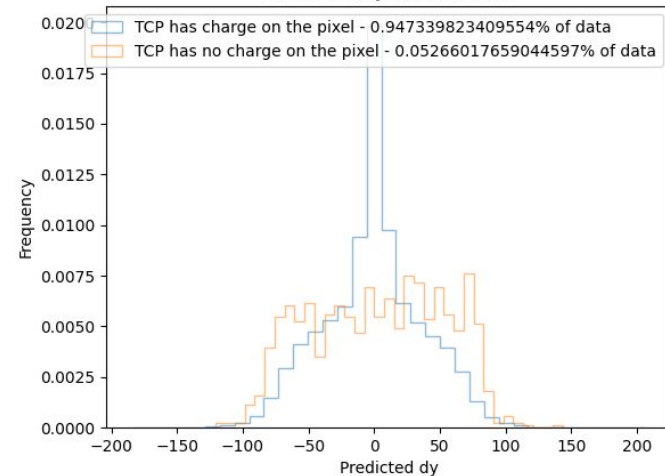
Pixel Window, layer 4



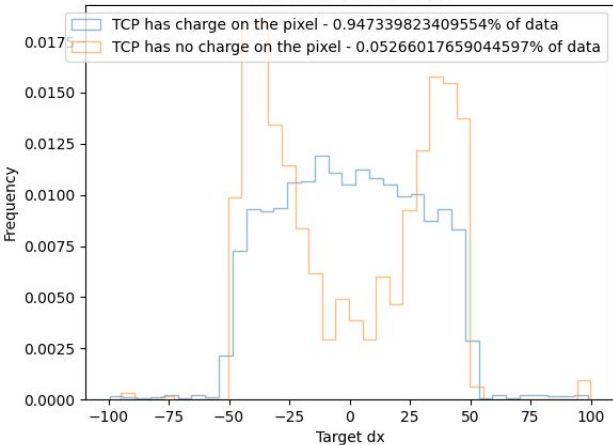
Predicted dx distributions



Predicted dy distributions



Target dx distributions



At the end, looked at if “chargeless” TCP’s were impacting - certainly have different distributions but only 5% of the TCP’s we get are chargeless

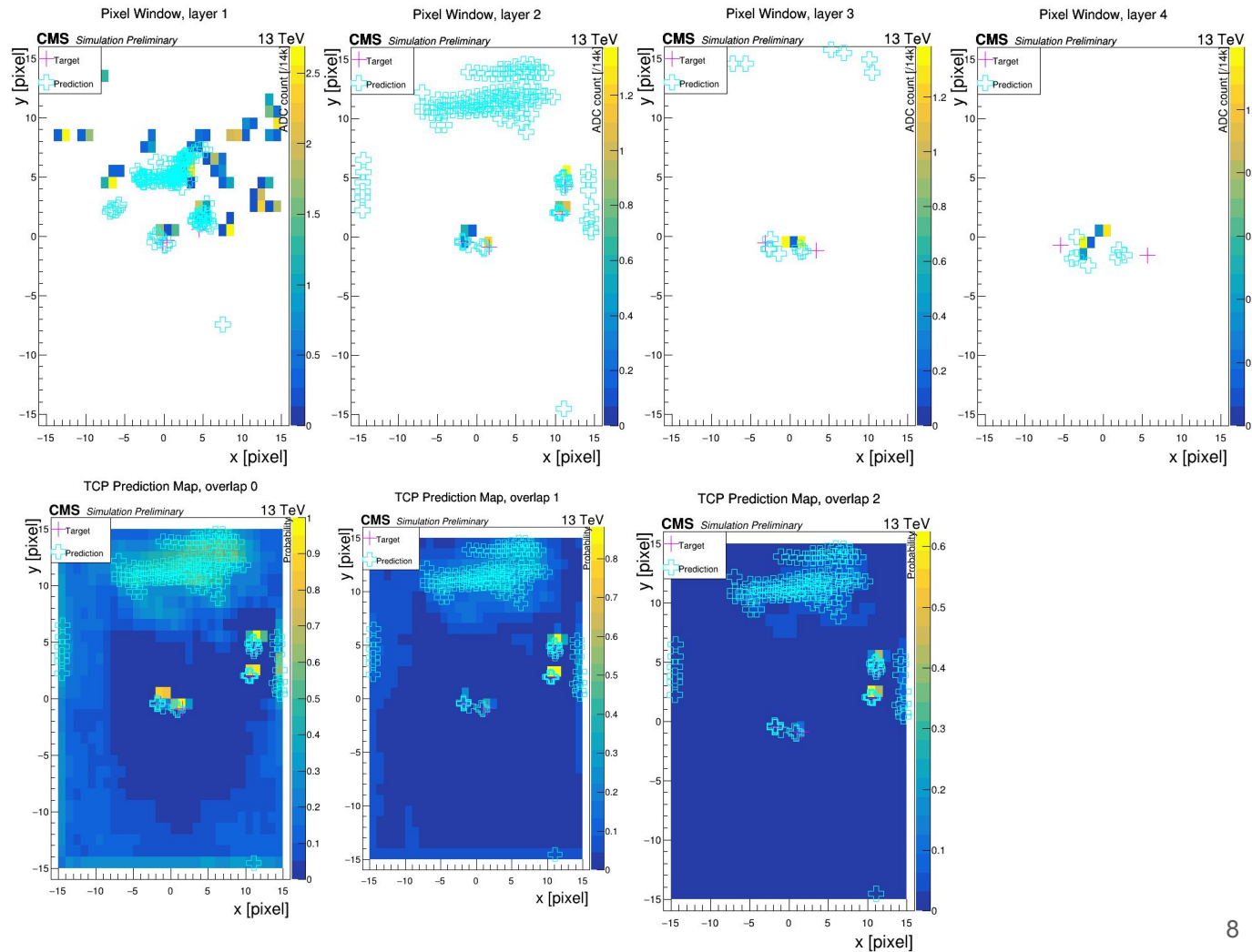
Current Tasks

- New notebook with more data/methods
- Investigate propagation between layers - how to improve on linear prop
- Visualizer checking

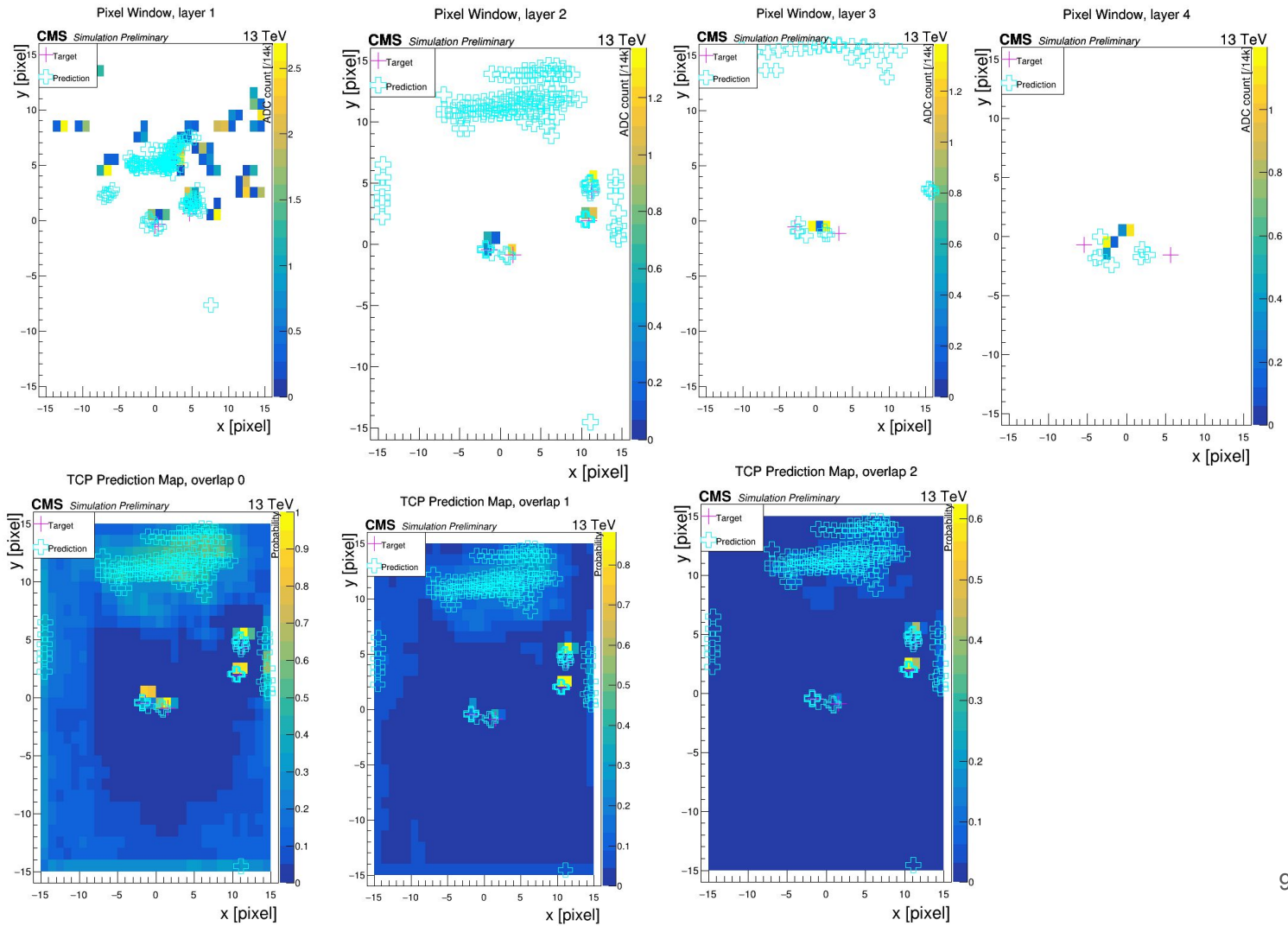
Linear Propagation

“The four barrel layers are radially located at $r_1 = 29$ mm, $r_2 = 68$ mm, $r_3 = 109$ mm, and $r_4 = 160$ mm” - the r_1 was set at 30, the r_3 was set at 102.
 “Pixel detector consists of 66M pixels ($100 \times 150 \mu\text{m}$)” - these values are correct for x and y distances of pixels

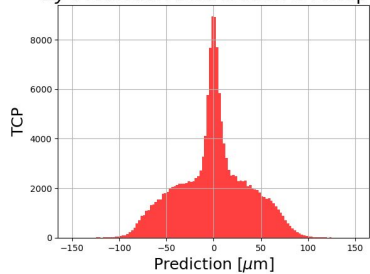
Plots with
new lin
prop



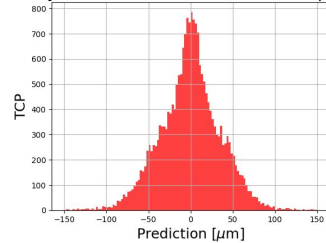
Plots with old
lin prop



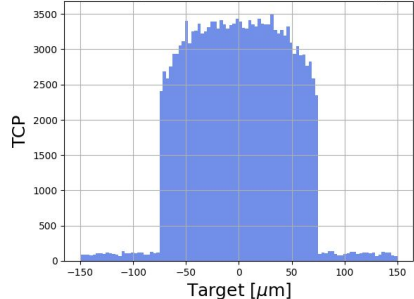
Δy Prediction Distribution Overlap 1



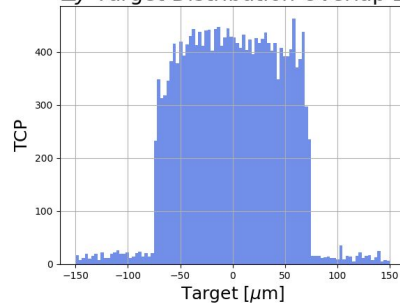
Δy Prediction Distribution Overlap 2



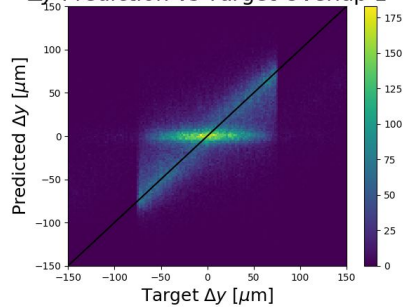
Δy Target Distribution Overlap 1



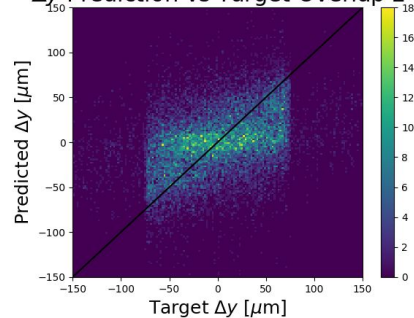
Δy Target Distribution Overlap 2



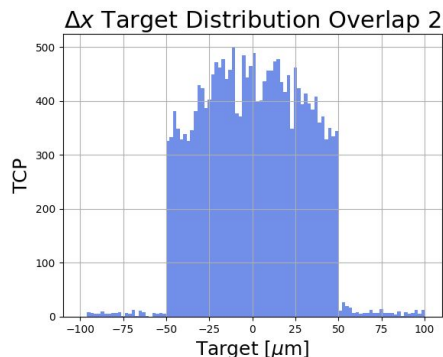
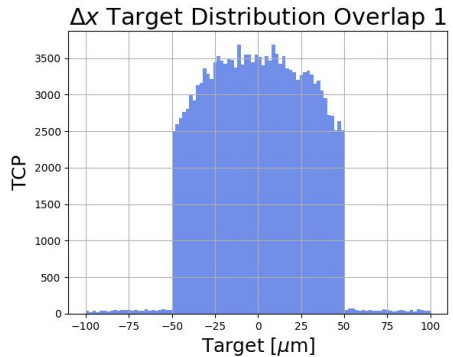
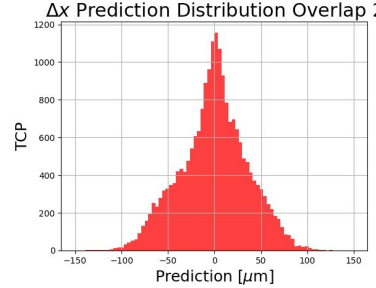
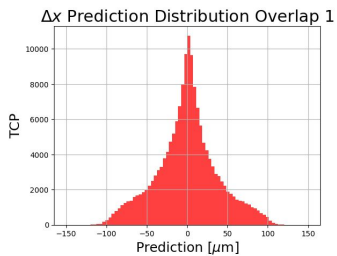
Δy Prediction vs Target Overlap 1



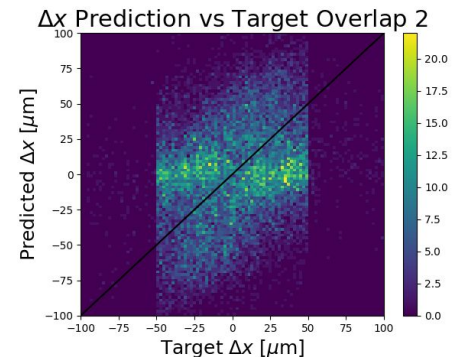
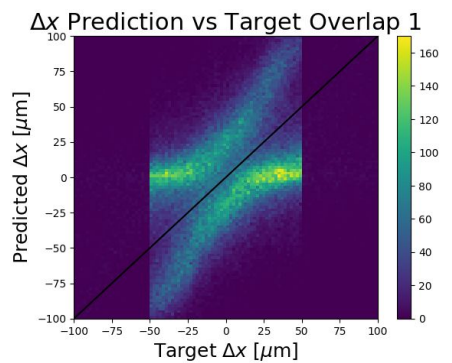
Δy Prediction vs Target Overlap 2



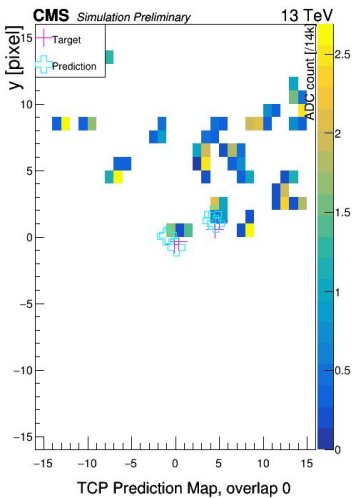
Targ + pred both
sent to 0 for
overlap 3



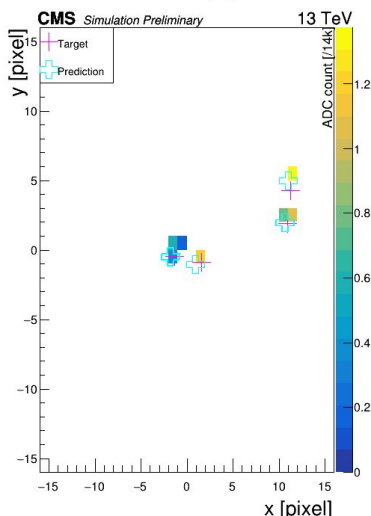
Targ + pred both
sent to 0 for overlap
3



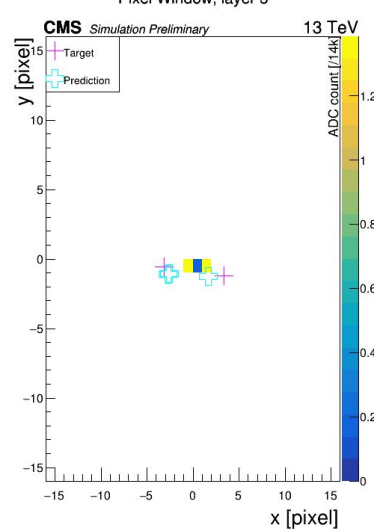
Pixel Window, layer 1



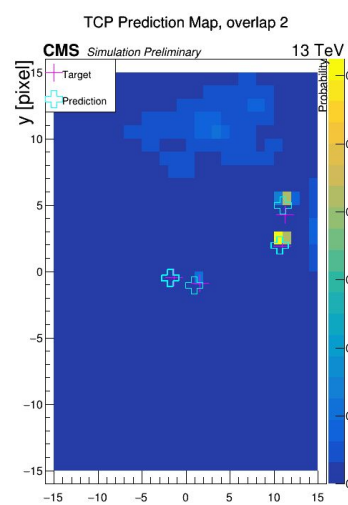
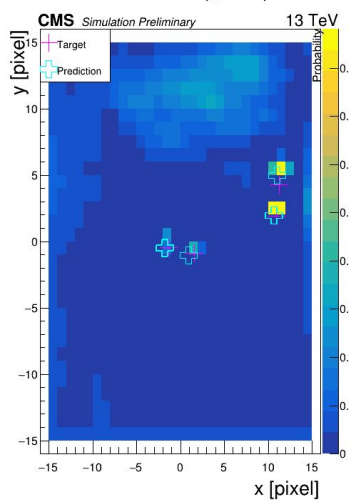
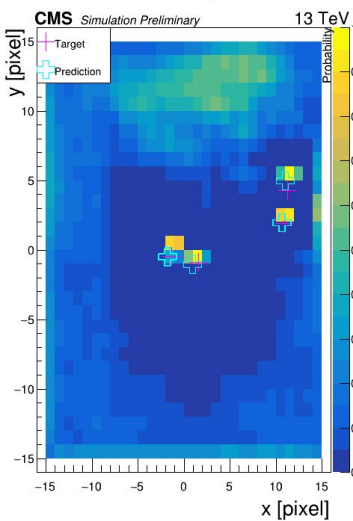
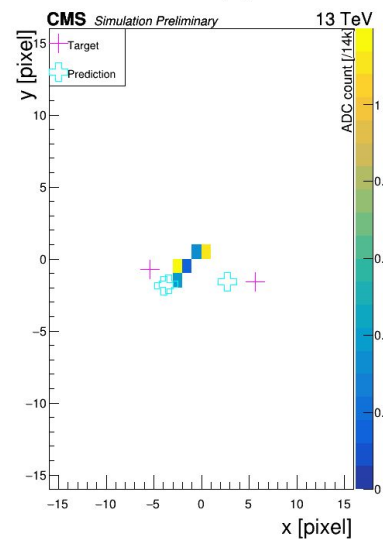
Pixel Window, layer 2



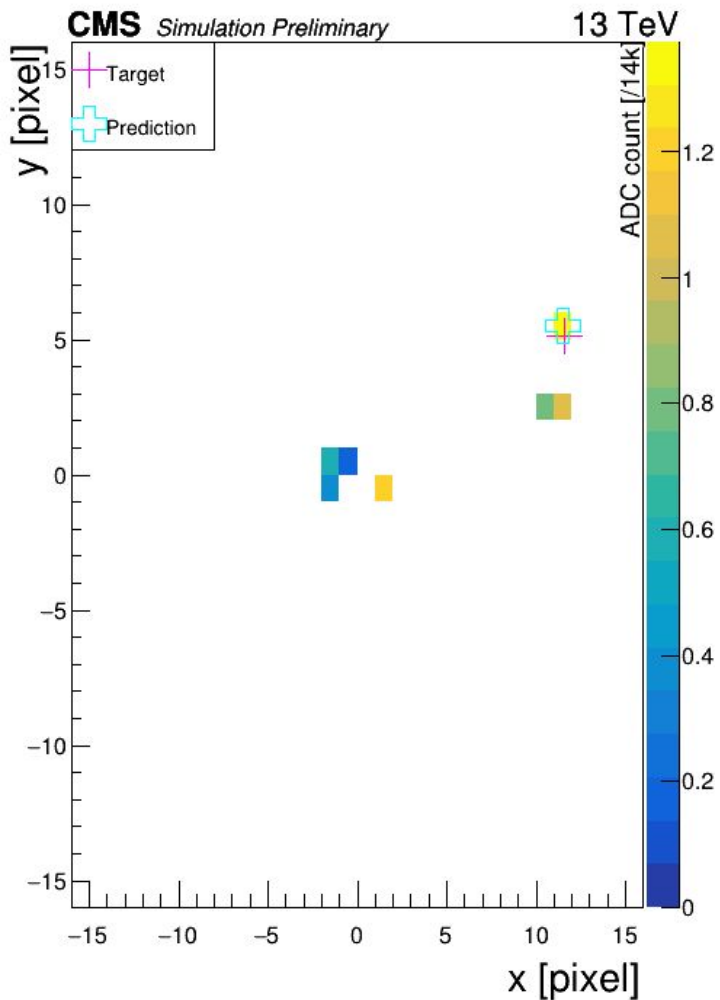
Pixel Window, layer 3



Pixel Window, layer 4



Pixel Window, layer 2

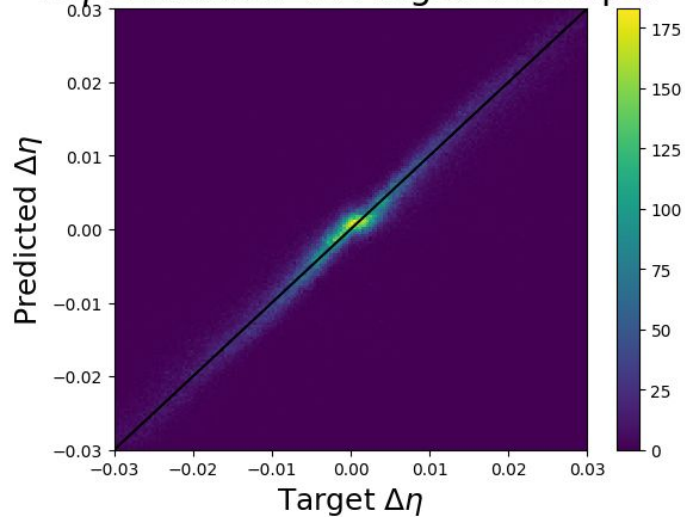


target(x,y,eta,phi)=
 0.24070096015930176
 -1.0722055435180664
 -1.1713344563576173
 -1.5955027470730188

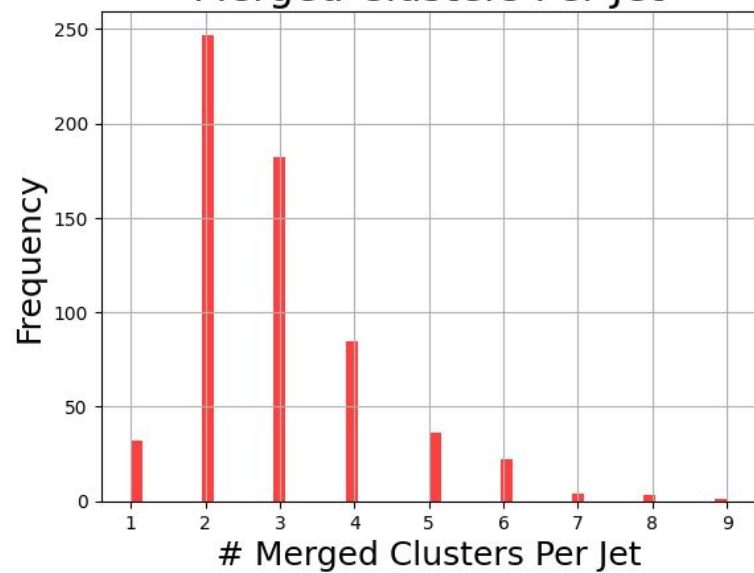
prediction(x,y,eta,phi)=
 0.028765248134732246
 -0.003075879067182541
 -1.2516465187072754
 -1.7314609289169312

(big pixel)

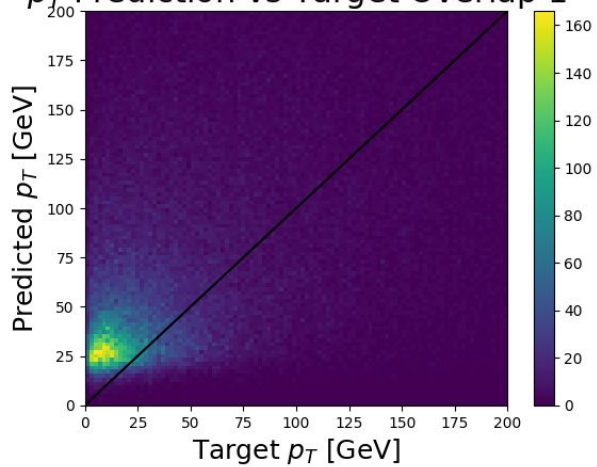
$\Delta\eta$ Prediction vs Target Overlap 1



Merged Clusters Per Jet



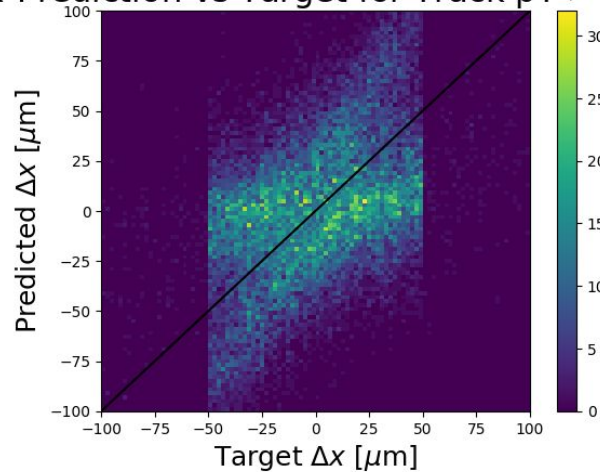
p_T Prediction vs Target Overlap 1



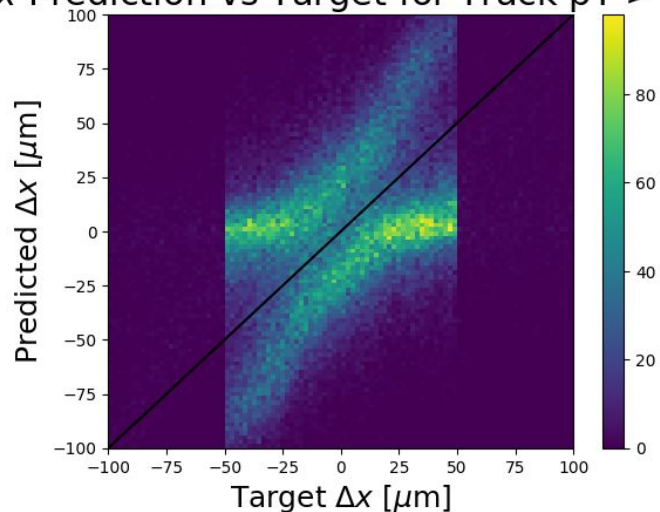
0.8125

1.0

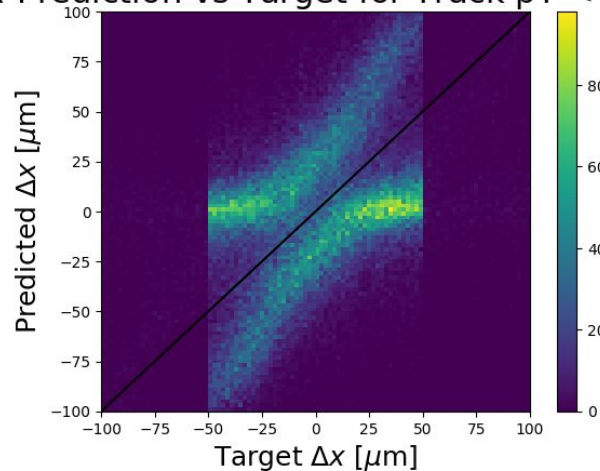
Δx Prediction vs Target for Track $p_T > 100$



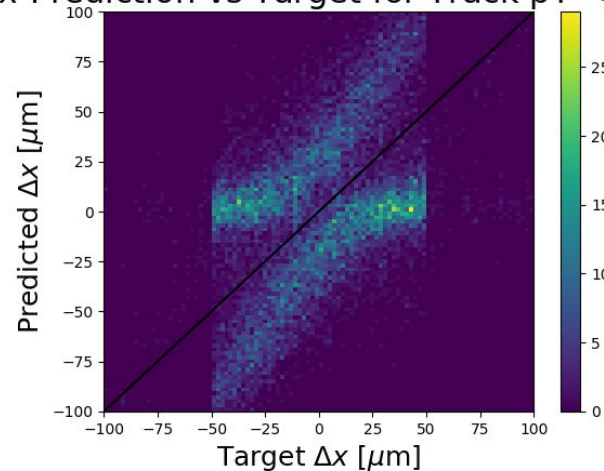
Δx Prediction vs Target for Track $p_T > 10$



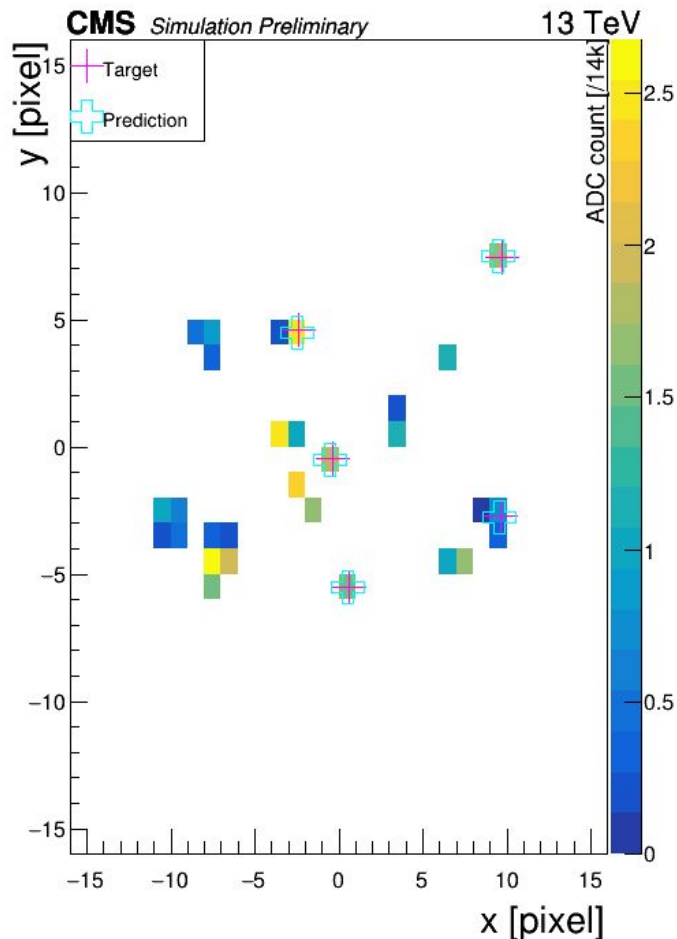
Δx Prediction vs Target for Track $p_T < 100$



Δx Prediction vs Target for Track $p_T < 10$



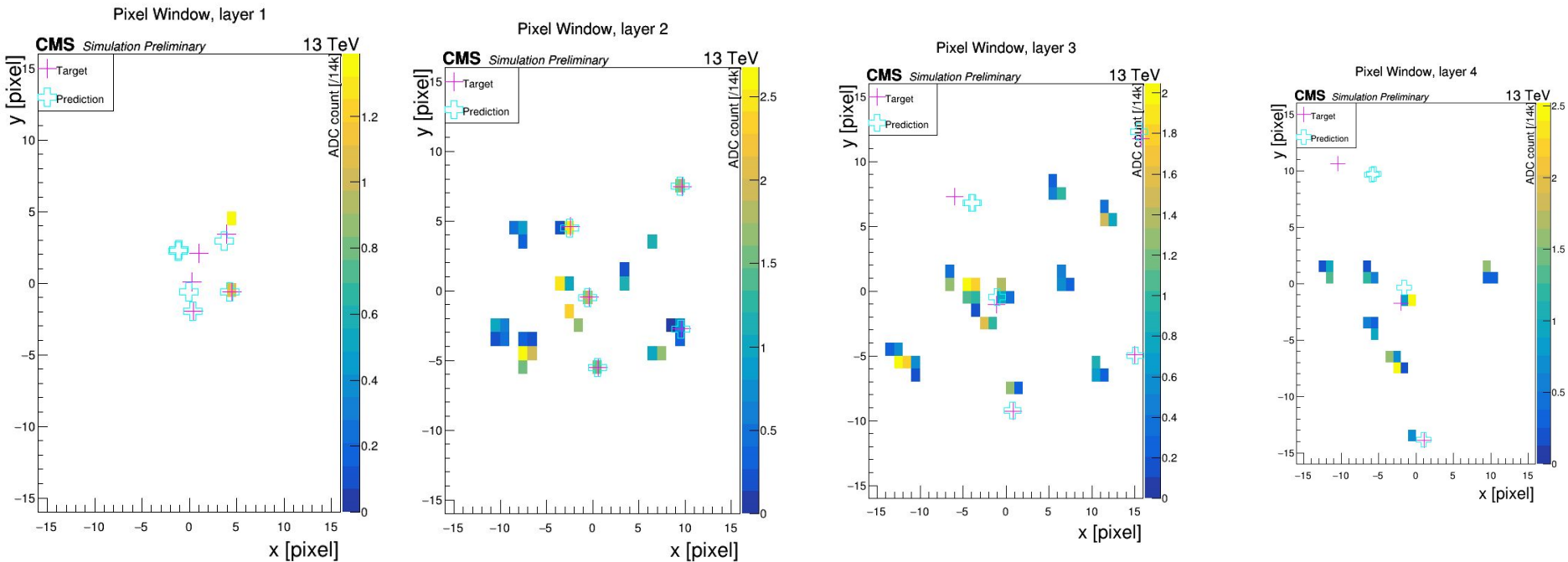
Pixel Window, layer 2



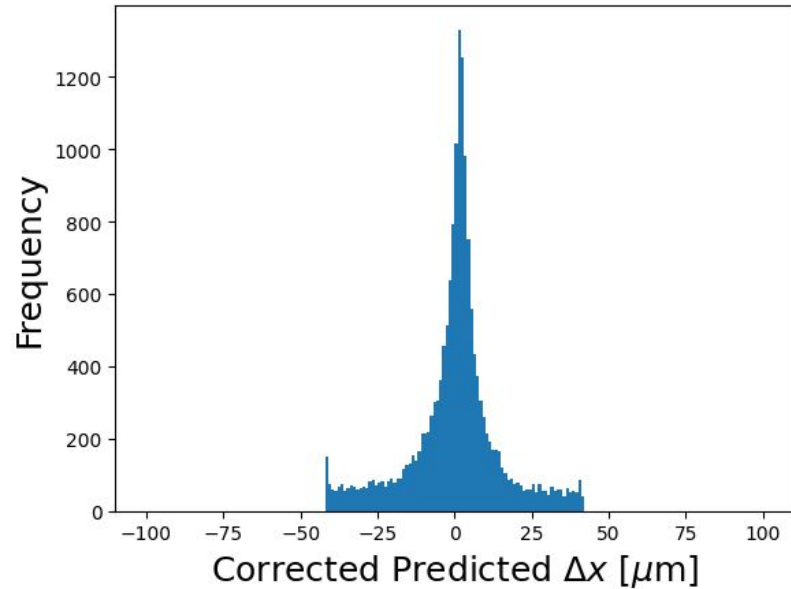
New Pred, bin (x,y): -3.0 4.0
target(x,y,eta,phi,pt)=
0.20136487483978271
0.355228066444397
-0.9074209666935618
0.8715723045369828
2.3911147441465905
prediction(x,y,eta,phi,pt)=
0.013919083401560783
-0.07324044406414032
-0.7944803237915039
0.3417768180370331
44.34525680541992

New Pred, bin (x,y): 0.0 -6.0
target(x,y,eta,phi,pt)=
0.22538554668426514
-0.05873811244964597
1.266343376705109
-0.05119968558644494
95.07100903743853
prediction(x,y,eta,phi,pt)=
0.08322035521268845
0.04299357533454895
1.2575995922088623
-0.05193641781806946
42.94028091430664

Add jet pt and jet eta



Add pred and targ dx, pt

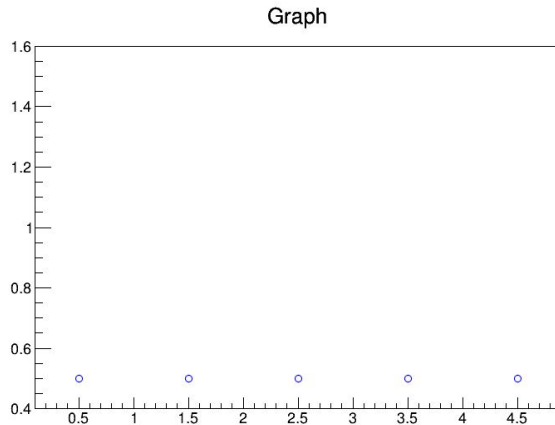


5th degree
polynomial fit -
extended range

Probability-based color gradient for prediction crosses

This is not working to
change the color of the
points - test case:

```
for i in range(num_points):  
    x = i + 0.5  
    y = 0.5  
    color = ROOT.kBlue + i  
    points.SetPoint(i, x, y)  
    points.SetMarkerStyle(20)  
    points.SetMarkerColor(color)  
points.Draw("AP")
```



Why are these all
the same color?

How CERN does it:

if (not ON_DATA) :

graphTargetTot[jet][1].Draw("SAME P")

graphPredTot[jet][1].Draw("SAME P")

ROOT forums say it is
impossible to do multiple colors
in the same graph, would need a
TMultiGraph/overlay graphs