

Position Resolution vs. Clock Speed in Simulations: Quick Results and looking forward

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Q-Pix Bi-weekly Meeting
(8/26/2020)

Overview

Since we are using QPix charge detection as a replacement for light detection for determining position, our goal is to use simulations to determine the position resolution as a function of clock precision.

Finding the minimum clock speed that doesn't affect resolution can tell us how to save on cost and power

We have done two things:

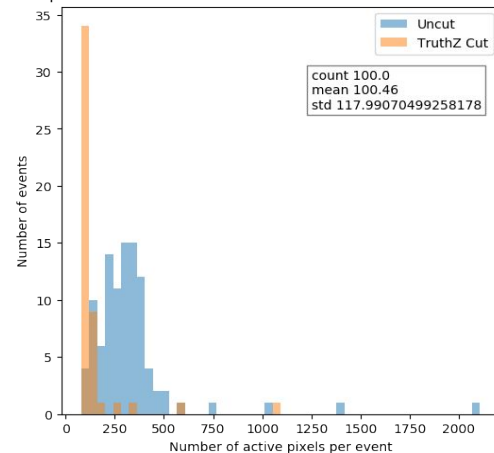
1. Long term: really understand the simulation data and get the resolution out of a cleaned data set
 - a. Many more effects than we were expecting. Data needs lots of cleanup from non-MIP events.
2. Short term (Since the summer is ending): Simple results

First quick comments on the data, then preliminary results

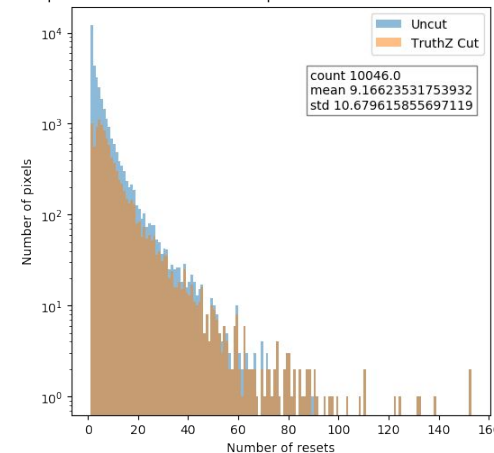
First look at the Sample

- Full sample of 100 simulated muon events
 - 32,688 active pixels
- Only want to look at pixels corresponding to an explicit drift distance (TruthZ) from the Geant4 data
 - 10,046 active pixels with TruthZ information

Comparison of Number of Active Pixels with and without TruthZ Cut

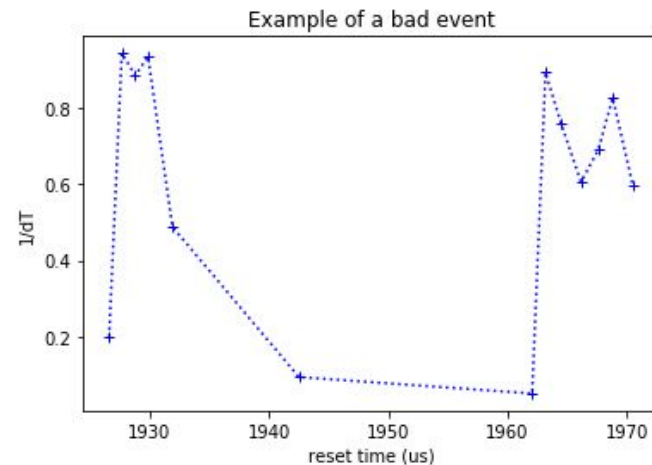
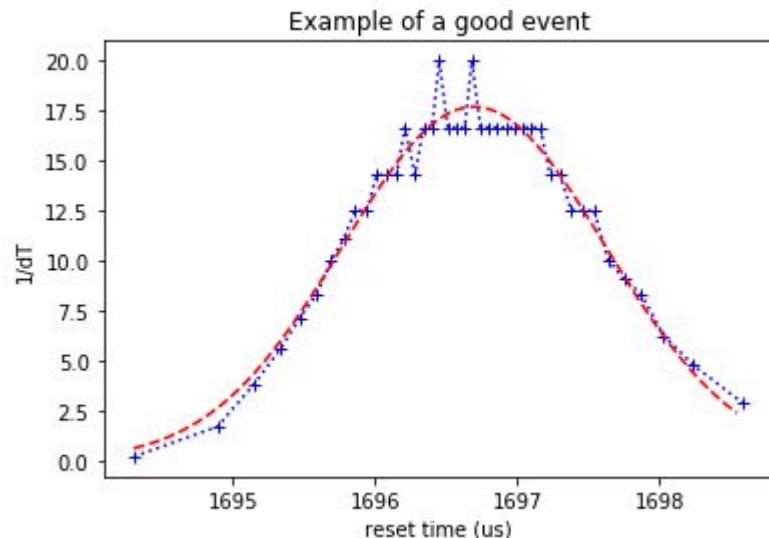


Comparison of Number of Resets per Pixel with and without TruthZ Cut



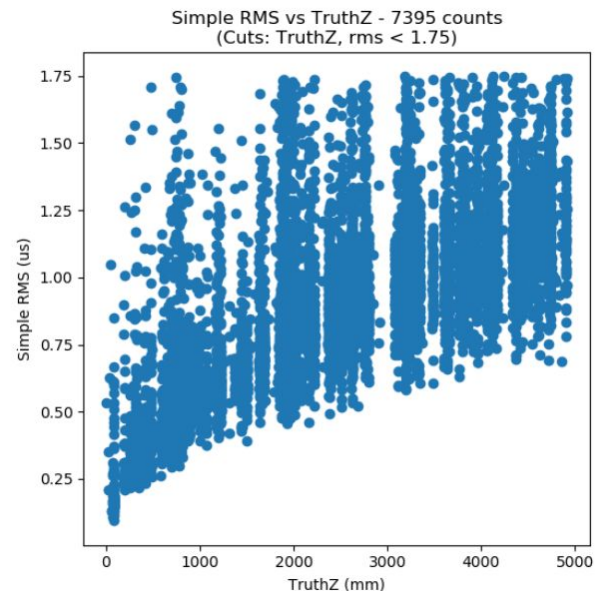
Good vs Bad Events

- We expect the diffused electron swarm produced from a MIP event to have a Gaussian density (current) both longitudinally and transversely in current as a function of time
 - Only care about longitudinal distribution for now
- Good (MIP-like) events will be Gaussian
- Bad events are not Gaussian or have multiple peaks
 - Still understanding their causes and developing methods to fix/remove them



Start with simple cleanup cuts

- Select a sample of MIP-like events with simple baseline cuts
 - TruthZ exists
 - nResets > 3
 - Simple RMS < 1.75
- Note that many events have larger than expected RMS because they don't look like MIPs

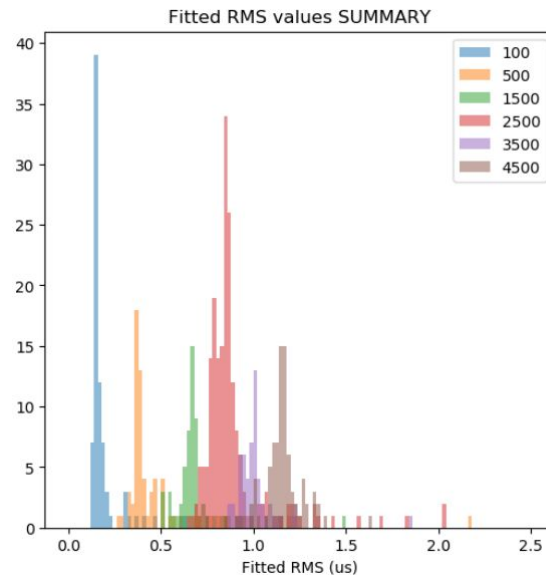


Fitted RMS for various TruthZ bins

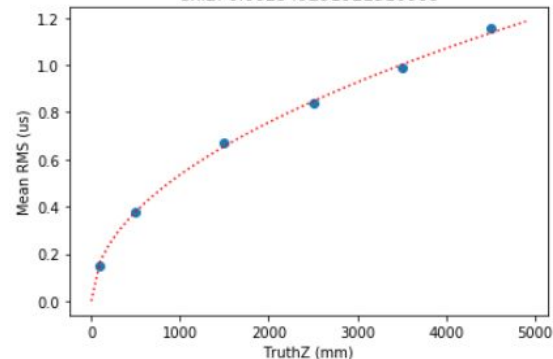
- After cleanup, fit the pixel and then group by TruthZ to determine the RMS vs. TruthZ relationship
 - If the pixel could not be fit, it was cut
- Consider events for a number of TruthZ distance - bin size of 100
- Fit each to a Gaussian in a restricted range to ignore long tails for simplicity
- Gives the expected sqrt RMS vs Z relationship
 - Resolution as a function of TruthZ is relatively flat (~ 0.04 - not shown here)
 - Can use this to determine clock speed recommendation

Ballpark uncertainty in Z at 2500 mm:

- We see an RMS of our measurement at about 0.04 on a mean of about 0.8. Call this about 5% measurement on RMS. Since Z prediction goes as RMS^2 , we get about a 10% prediction of Z which is about 250 mm



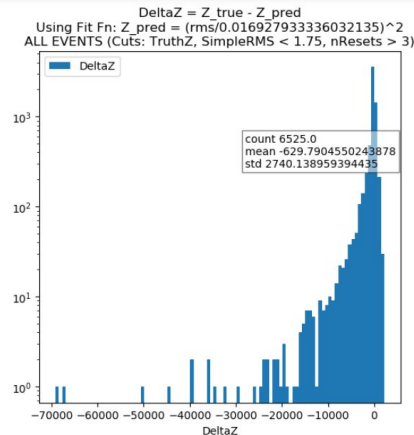
Fitted Mean RMS vs TrueZ based on samples at 100,500,1500,2500... ± 50
Fit Function: $y = 0.016927933336032135 \cdot \sqrt{x}$
Chi2: 0.002549291921510066



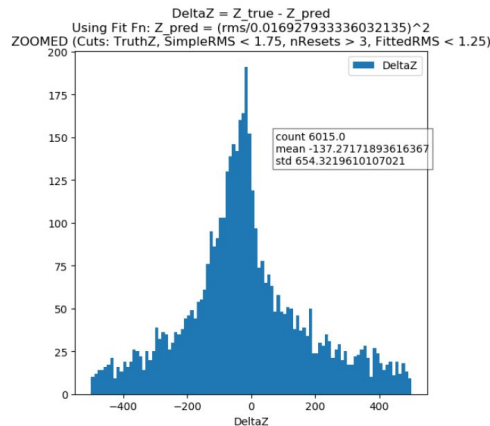
Long run strategy:

- *How well do we measure the Z value using the simple event selection methods?*
 - Determined measured_Z from fitted RMS of the pixel
 - Right: $\Delta Z = \text{true_Z} - \text{measured_Z}$
- Dominated by long tails of mis-measured events. Will take more study
- RMS in the peak is about ~150mm
 - Can be compared to 250mm (expected to change as a function of TruthZ)
 - 150mm isn't a fair estimate since we are summing Gaussians with different resolutions

Full range of events



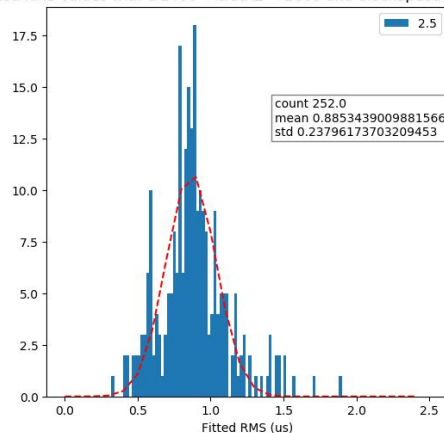
Zoomed to [-500,500]



Simple method for determining resolution as a function of clock speed: Pick a single point in Z

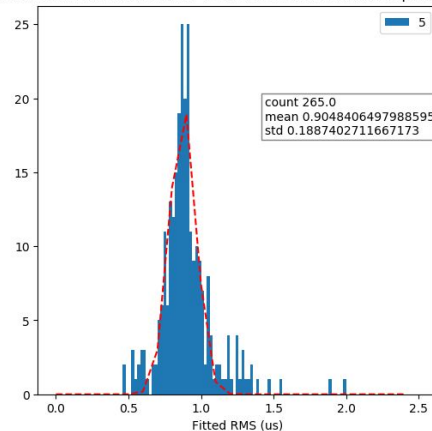
Expand range from $2400 < \text{TruthZ} < 2600$ to get higher statistics, and look at distribution for different clock speeds. Resolution gets worse for slower clock speeds.

Fitted RMS values with a $2400 < \text{truthZ} < 2600$ and ClockSpeed = 2.5MHz



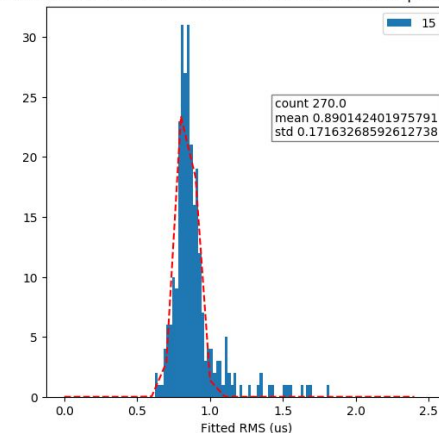
```
a = 10.84853814429051 ---- unc: 0.883593196008633
mean = 0.866163916032286 ---- unc: 0.01626708664762082
var = 0.029138870516594816 ---- unc: 0.005715736371260217
[21, 0.87, 0.28891739354978613]
<class 'function'> <class 'list'> <class 'list'> <class 'list'>
```

Fitted RMS values with a $2400 < \text{truthZ} < 2600$ and ClockSpeed = 5MHz



```
a = 19.687001389904722 ---- unc: 1.1566035734286602
mean = 0.8745658583021488 ---- unc: 0.0060995999534144444
var = 0.008088987222698147 ---- unc: 0.0011045047107235602
[23, 0.87, 0.2344687735824125]
<class 'function'> <class 'list'> <class 'list'> <class 'list'>
```

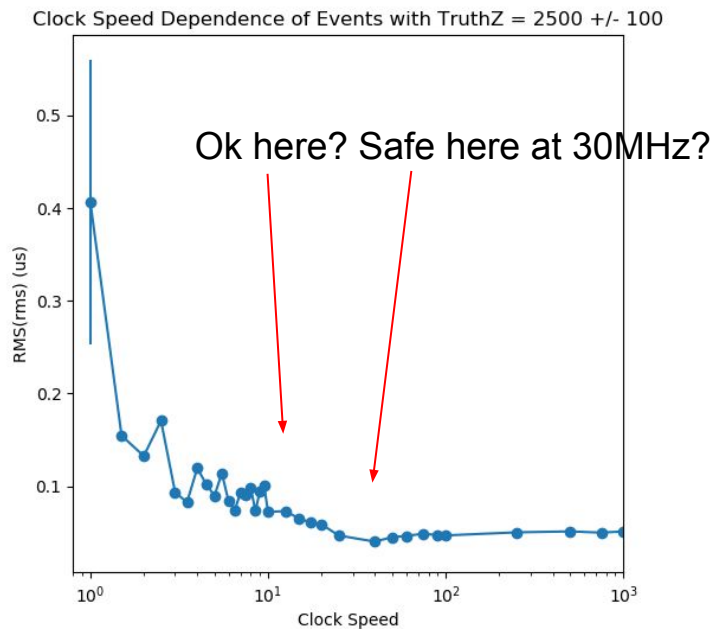
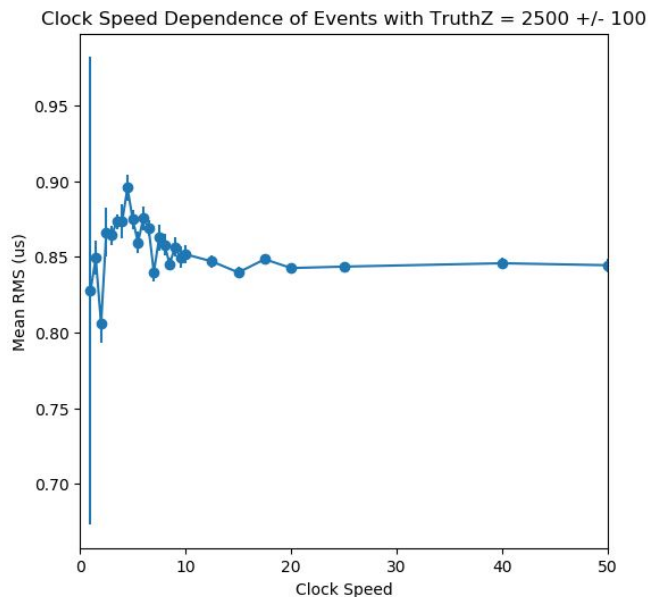
Fitted RMS values with a $2400 < \text{truthZ} < 2600$ and ClockSpeed = 15MHz



```
a = 28.125635646852785 ---- unc: 1.2973097036523538
mean = 0.8396019868769987 ---- unc: 0.0034744302958493404
var = 0.004256006491470612 ---- unc: 0.0004533671008773532
[32, 0.87, 0.18621015942099448]
<class 'function'> <class 'list'> <class 'list'> <class 'list'>
```


Results

Best fit measured $_Z$ stays about the same (as expected), and resolution (width of the RMS distribution) gets worse for slower clock speeds. More statistics would smooth this out.



Conclusions? Next steps

- Looks like we can get an expected resolution of about ~200mm but that it should be Z dependent
- A 30MHz clock speed appears safe. Getting a more robust number will take more work
- Many events still don't have the correct Z measurement due to non-Gaussian charge collection. Cleaning up the sample is our next step