# Making The StDev\_Min Method More Robust

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### Outline

- Recap: the StDev\_Min Method and Known Pathologies
- Addressing Type 1 Pathologies: Delta\_StDev Method
  - Simple attempt w/ t0=0 assumption
  - Iterative process to remove t0=0 assumption
- Addressing Type 2 Pathologies: Decreasing Reset Threshold
- Wrapping it up: Next Steps, Thoughts and Conclusions

### Recap

#### The StDev\_Min Method

Previously introduced the StDev\_Min method:

For every event:

- Look at all pixels in an event with nResets  $\geq 8$
- Find the pixel with the minimum standard deviation of the reset times (since StDev only fluctuates low in a systematic way)
- Use the StDev\_Min pixel from each event to get the functional form of Z vs StDev\_Min
- Use the functional form to calculate the Z and  $t_0$  of other pixels
- Use the StDev\_Min pixel as the basis for the rest of the event reconstruction



#### Two issues with the StDev\_Min Method

The method is a very simple first try and there are two known reasons as to why the smallest StDev may not lead to a good measure of t0:

**Type 1 Pathologies)** Algorithm is susceptible to outlier pixels (*see top right*). If a particle branches off and deposits its energy at a much smaller Z than the rest of the event, there is a good chance that the algorithm will pick that pixel regardless of whether there is a cleaner pixel readout.

**Type 2 Pathologies)** Algorithm is susceptible to low statistics (*see bottom right*). Because of the dE/dx nature of the physics, there are very few resets that occur from a single clean track. 8 resets in a pixel readout usually only happens from multiple interactions. Therefore the minimum StDev of 8+ resets tends to be higher than the predicted StDev for the corresponding Z. When there are very few 8+ reset pixels, the chance that one fluctuated low (clean) decreases.



### Addressing Type 1 Pathologies

# Tweak the results of StDev vs TOA functional form before using it

Remove outliers contributing more than 3 sigma to chi squared that biased the functional form and refit. Doesn't change much.

Three plots shown here:

- Red is the theoretical relation
- Blue is the original fit (with all 100 points)
- Green is new fit (excluding the solid blue outliers)

```
StDev_Min = 1.9872e-05*sqrt(Mean_TOA)
```



#### More robust method of picking the correct pixel

Since we want fluctuations low, we have to be more thoughtful about picking the pixel with the smallest value.

- Move to using the variable which is the difference between the *expected Stdev for this time*, and the observed Stdev
  - Use the "fluctuated\_low" functional form for comparison
- Pick the pixel with DeltaStDev\_Min
- Will fix the bad assumption that t\_0=0 soon



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#### Results after moving to the new method

Determine the *true* StDev\_Min to be the pixel from each event that has the most negative (smallest) Delta\_StDev compared to the predicted value given by the functional form at the given Mean\_TOA.

- **Green x's** (highlighted by green arrows) show type 1 pathologies that were removed with the DeltaStDev\_Min method
  - As expected, due to the nature of Type 1 Pathologies, these only occur in the smaller TOA section of the plot
- Blue are the two blue outlier events shown in the bottom right hand plot
- Orange are non-outlier points (contributing less than 3 sigma to chi squared)

#### Revised fit is now:

StDev\_Min = 1.9852e-05\*sqrt(Mean\_TOA)



### Calculating t<sub>0</sub> from TOA

- Distribution is Gaussian, but is shifted from zero as well as has tails and outliers
- <u>Most</u> of the variation in this plot comes from the variation in Delta\_StDev
  - If we could find a way to reduce the variation in Delta\_StDev, we could clean this up
  - Still biased to as a function of Z, but we'll deal with this after we finish the next section
- The tails come from the fact that the functional form has small biases as a function of Z, and the resolution is also rising as a function of Z.



#### Plotting Delta\_TOA vs StDev\_Min



- Still has a Z dependence
- Has a resolution of about 100-150us at Z=300cm 11

#### Correcting the t0=0 Assumption

The previous algorithm only works if we have the right x-axis, and in simulation we have the unfair advantage of t0=0 for all events

Can iterate:

- Start with the pixel with the smallest StDev
- Assume that the StDev corresponds to the yellow line fit
- Then, re-look at all StDev's and find the minimum DeltaStDev and assume that is on the yellow line. This is a direct best-estimate of the t0

#### Step-by-Step Iteration of Minimizing DeltaStDev Using E80

Step 1: Pick the pixel with smallest StDev



Step 2: Assume the pixel corresponds to the yellow fit line

A pixel that falls on the fit line with a StDev of 0.492us corresponds to a Mean\_TOA of 614.5us

Since it actually has a Mean\_TOA of 331us, we need to adjust all Mean\_TOAs by adding 614.5-331 = 283.5us and calculate DeltaStDev based off their shifted positions and the yellow fit line Step 3: Calculate DeltaStDev's and re-evaluate. Put new DeltaStDev\_Min on yellow line



#### Step-by-Step Iteration of Minimizing DeltaStDev Using E80

Repeating the process, except this time, using Pixel [291, 993]:

Mean\_TOA = 980us StDev = 0.621us



A StDev of 0.621us corresponds to a TOA of 980.1us  $\rightarrow$  All Mean\_TOAs will be shifted 0.1us.



...Therefore, this confirms we should use Pixel [291, 993] in  $t_0$  calculations, and it successfully measures to to 0.1us.

#### Results of TOA\_Shifting (without t0=0 assumption)



- All TOA\_Shifting data matches up with the DeltaStDev data, as expected
- Now, we are getting the same correct answer, but without working on the assumption that we are starting with the correct Mean\_TOA

#### Results of TOA\_Shifting (without t0=0 assumption)



We can see that the results are almost identical, and we arrived at these results WITHOUT assuming t0 = 0. This means that we should be able to get a good measure on t0 using this method on any arbitrary t0 and TOA

### Addressing Type 2 Pathologies

#### Change 2: Decreasing Reset Threshold

**Problem to solve:** In order to get enough resets to dependably measure the StDev, we are limited to looking at mostly multi-deposit pixel readouts, keeping us from ever getting many clean measurement

**Solution Concept:** Lowering the reset threshold will produce more resets per pixel. For example, if we lower the reset threshold by 50% (From 1 reset = 6250e<sup>-</sup> to 1 reset = 3125e<sup>-</sup>), we can get twice the resets out of a deposit and start considering more pixels in each event.

**Expected Benefit:** More resets per pixel raises the probability that a clean deposit has 8 resets. This should improve the resolution as well as reduce events that fluctuate to small numbers of pixels (and t0 mismeasurements)

## Decreasing the reset threshold increases the number of pixels with 8+ resets per event



	Factor decrease of reset threshold	Factor increase of 8+ reset pixels	Percentage of new pixels added
1fc	_	_	_
0.5fc	2	8.3	25%
0.375fc	2.67	15.6	32%
0.33fc	3	17.8	35%

- With the same sample, we can drastically increase the number of 8+ reset pixels while only adding a modest number of new pixels.
- This also allows us to measure pixels that were formerly only 3+ reset pixels. This means that the chances of finding a clean pixel are much higher.

# Reset times for the same pixel with two different reset thresholds



In this event, by decreasing the reset threshold by 50%, it increases the number of resets by 100%. The pixel shown above would not have been considered under the original reset threshold, but it (along with  $\sim$ 700 more pixels per event) can now be used to find a good t<sub>0</sub> measurement. Pixels that were originally able to be measured can now be measured with better resolution.

# Results with New Reset Thresholds

Orange points are the points using DeltaStDev\_Min method.

Can see clearly that the decreasing thresholds pull the points closer to the fit line.

We also note that the fit line moves slightly lower, as expected.



# Showing Resolution and Systematic Bias

The resolution is improving as a function of reset threshold.

The systematic bias as a function of Z also goes away and becomes flat within statistics.

(Will compare all on one plot in the next slide)







#### Comparing T0 Resolution as function of reset threshold New Sample: 100 events at Z=300cm (for better statistics)



### Wrapping It Up: Next Steps, and Conclusions

#### Conclusions

- The simple StDev\_Min solution was a good start, but was susceptible to 2 types of pathologies leading to misleading results
- Using the DeltaStDev\_Min method remedies the type 1 pathologies, while still remaining open to clean outlier pixels, and also removes the t0=0 assumption
- Decreasing the reset threshold has big impact on the resolution of to calculations.
  - Decreasing the threshold greatly increases the number of pixels with nResets  $\geq 8$
- While there are many more checks that could be done, using both methods together allows for a dependable measurement of t0 on the order of 10us or less
- Next step is to turn this into a simple reconstruction algorithm and release