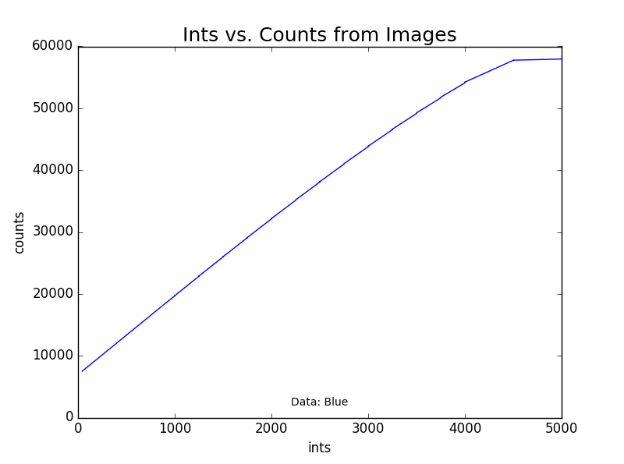
Clio Linearity Calibration: Draft

Chris Bohlman

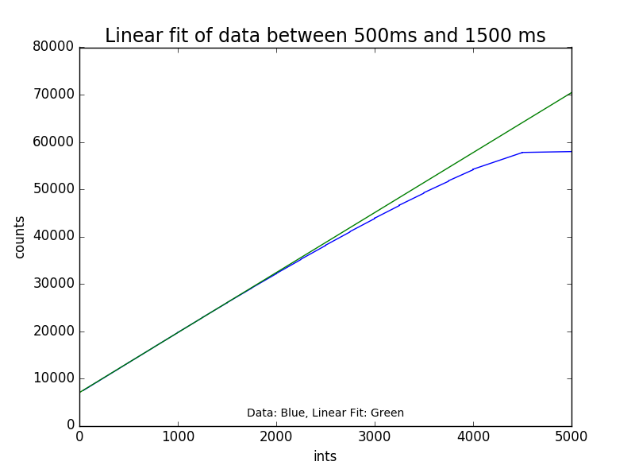
10/25/2016

Ints vs. Counts:



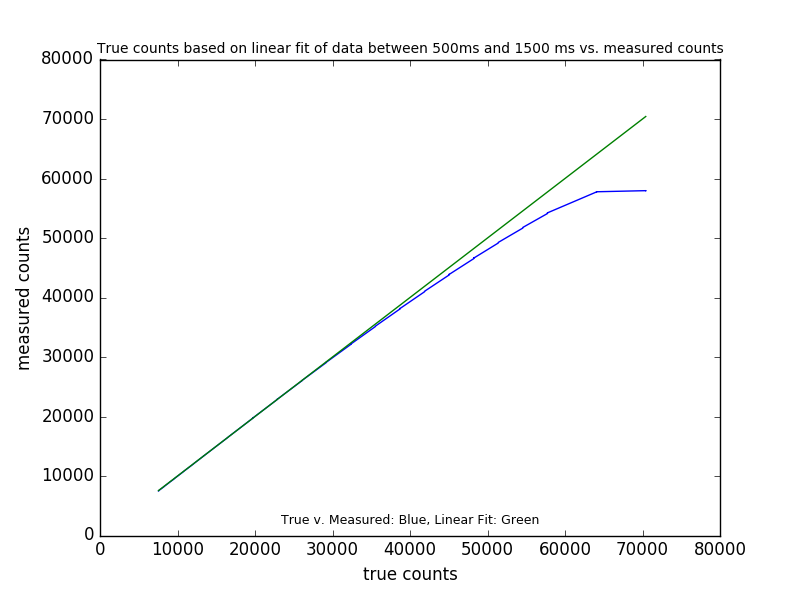
First started by reading in each of the files through fits and, when given a certain area of the picture to analyze, analyze the number of counts in said section of the picture, and find the integration time of the picture in the picture details. This was done through all of the given files, and the data was stored in 2 separated arrays, appropriately called ‘ints’ and ‘counts’. When these two arrays were graphed against each other, the outcome was the graph above.

Ints vs. Counts with linear relationship:



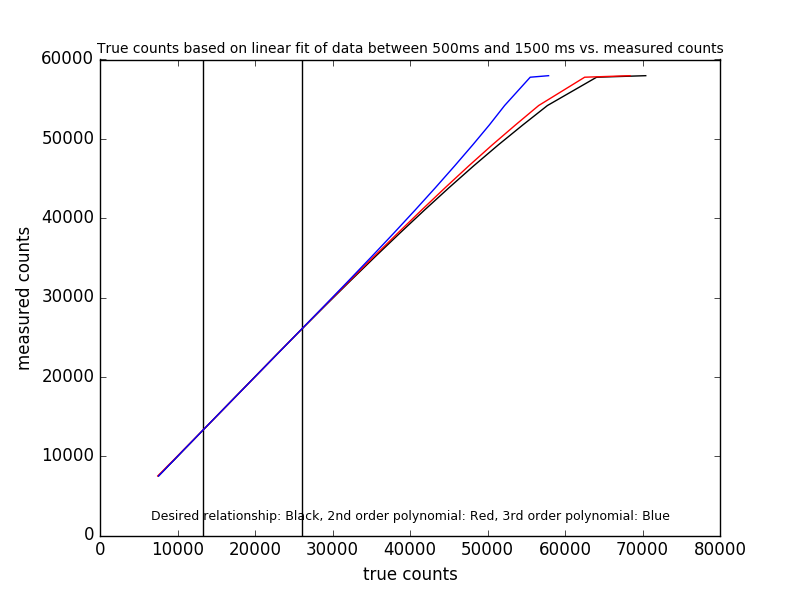
I needed to find a linear relationship with the data so I could find a method of determining the true counts, so I decided that the most linear part of the graph was between 500 ints and 1500 ints. Since every 5 ints was the same, along with slightly different counts for each int in the group, I used polyfit from numpy to determine the coefficients of a line analogous with the area that I had picked out. I then plotted said line next to the data, showing where the data is linear and where it diverges from my calculated line.

True counts vs. Measured Counts with linear:



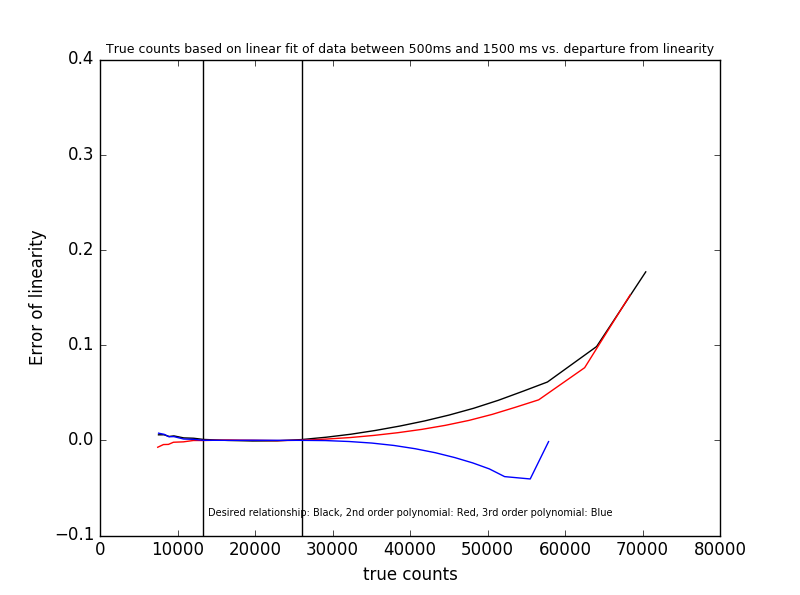
To determine the error of the counts as the measurement approached a higher integration time, a method of determining the true amount of counts (if the counts did not lessen) was needed. Fortunately, that was relatively simple to do with a linear approximation of the same part of the graph as before. I again used polyfit to formulate a linear approximation of the line between 500 and 1500 ints, but instead of directly graphing it onto the graph again, I evaluated the coefficients given by treating it as a polynomial. Because polyfit gives the ‘A’ and ‘B’ coefficients in ‘Ax + B’ when evaluating the section linearly, I made a true counts list by appending the polyfit polynomial evaluated with every int from 0 to the 114th int. The result was a list of true counts that was the same length as the ints list and the counts list. To graph it on the same graph as the original data, I first graphed true counts vs. true counts to give a line of comparison, and then I graphed true counts vs. the original counts to model the data.

True counts vs. Measured Counts with all:



In the end, what was wanted was to see which order equation could model most of the data the best. Therefore, once I had the linear version of the true counts, I calculated the coefficients of a second order equation per the data, and a third order equation to the data. When getting up to higher orders, it was important that I modify the lists of ints/counts a little bit. Because all the ints were the same in groups of 5, and all the coefficients in this group were slightly different, for every int, I averaged out the amount of counts across the five counts for a single int, and made a list of every int without repeats and the subsequent averaged out count. I then used polyfit on these to model second and third order relationships, and redid the first order linear, since that was going to be my desired relationship of data. I then graphed true counts vs counts for every single order the same way I did it in the method of above, and graphed them on the same graph as each other. The vertical lines are representations of where the coefficients for each of the lines were gathered from.

Error of Linearity vs. True Counts



In the end, we wanted to see which order (2nd or 3rd) was closest to the desired relationship of first order. Therefore, stemming from the last graph, I produced an error of linearity plot. Now the true counts axis stayed the same, but for the error of linearity, I used the equation

*Error of linearity* = (*True Counts – Measured Counts)/True Counts*

For every single count, I calculated the error of linearity and stuck that into a list. I then plotted the error of linearity compared to the true counts for first order, second order, and third order, and analyzed the results based off how close the second and third order errors were to the desired relationship of first order.

Analyzing the Data:

From the error of linearity plot, I discovered that the second order equation behaved more closely to the desired relationship. Now, the reason I couldn’t add fourth order is because there were only 23 items on each list (due to me averaging the ints and counts into separate lists), so the fourth order polyfit had some odd behavior. After consolidating my data and limits and methods across three python files, I finally came up with that error of linearity graph, which was a marked improvement over what I was producing previously. However, because I couldn’t get any order to exactly match up to the desired relationship, I was at a loss for things to do. That’s why I am writing this report and organizing all my files: to get some ideas or something, or to notice a method that I did wrong.

Here’s to hoping that happens.