Broad structure of the code:

Part#1: Generating correlations S_{mn}

class treeProcessor →

- takes care of opening the .root file containing the eventbuilt tree and drawing things from it. Also maintains correlation histograms and saves them.
- everything SPS+SABRE related happens here, gainmatching, identifying channels, all else
- in the scheme of analysis, an object of this class is asked to do the following:
 - open the ROOT file corresponding to the right run number
 - be ready to look through all its 'events' in coincidence when asked: we only care about coincident MMM ring+wedge for a single MMM petal of SABRE('board' from here on, numbered 0 to 4)
 - have ready the channelmap, have ready the ability to zero in on a given SABRE detector's data when asked for it
 - jump through the data when asked for by a LOOP (more on that below) and report all events that have a particular (ring, wedge) pair fire with non zero energy in both, and single multiplicity
 - keep filling the ring, wedge energies in a scatterplot named 'scatter' as you go along

THE LOOP:

- lives in main.cpp
- Asks the treeProcessor to jump through the eventbuilder output as described above, for a particular board's particular (Ring, Wedge) pair
- For each coincident, 'good' (ring, wedge) event found by the treeProcessor, send the ring, wedge energies (A_m, A_n) to the likelihoodNet.

class likelihoodNet

- The likelihoodNet class keeps an array lkGrid. The x position of each point in lkGrid (the array index) corresponds to a particular value of the slope S_{mn}. What's stored in each array position is the likelihood that that slope is the real E correlation between the (ring,wedge) pair at hand
- The lkGrid, hence, is the probability distribution for each likelihood in the range chosen. We start out with the prior that all likelihoods in the range Smin, Smax are equally likely and see how as we step through data that assumption evolves. Eventually, we expect to have a peaked distribution that gives us a maximum, and a measure of variance.
- NOTE: likelihood based calculations are meaningful even when data is sparse. The paper assumes the slope to have the same distribution as that of a random variable that's the ratio of two gaussian distributed random variables A_m and A_n the Cauchy-Lorentz distribution.
- Robust linear fits to data are also obtainable by using the option ROB=0.x in ROOT fits. We do both of these here.

S = getGridXAt(i);

- The above two lines perform the bayesian propagation described in eqs 7.8,9 in Reese(2015). We then normalize the distribution as we step forward (divide every lkGrid element by 1/sum of the grid)
- After we've stepped through every instance where there's energies in the (ring, wedge) pair chosen – likelihoodNet can then be integrated over by summing the normalized distribution. When it crosses the 16% and 84% threshold, we get the uncertainties about the median. 50% point is the median – we avoid mean in the interest of robustness. Between 16 and 84 percentile lies 68% of the data – which gives us our S_{mn} with error.
- The executable here is called 'findCorrelations' and takes as argument \$runID \$inputfile \$r \$w
- ./correlateAllMMMs.sh writes the output the executable to text files titled outlist.0, outlist.1 ... outlist.4 that live in the directory runXXX/det0, runXXX/det1 etc respectively
- These are formatted clearly enough so as to be readable, it only makes the program in #3 sweat a little extra to clean up the formatting and extract the slope and error from it.
- Lastly, the output from doing all these correlations is also written as png files to the folder that looks like runXXX/det[0-4]/test_detector[0-4]_rng\$r_wdg\$w.png this way one can verify that the slopes have reasonable agreement with correlations seen.

Part#2: Find the global scaling factors we will need, from Am-241 data

- What's this?
- When we perform gainmatching this way, we ultimately get all factors s_n such that we can convert ADC readout A_n on channel n to energy E as $E = s_n A_n$
- Thing to note here is that all s_n 's could be scaled by the same factor α and nothing would change the calibration still holds.
- It helps us out quite a bit if we set this α such that 1keV = 1channel holds for SABRE data. In this step, we basically scour some raw ROOT data stored with Am-241 peaks in it, that can help us find this scale factors.
- calibrator.cpp has the main() function inside it, since it's so trivial. I decided to write c++ for this just so the entire project is consistent
- Basically, we read a tree, identify the back-most SABRE rings, and scale the E peak we see such that it falls on 5486 channels. We get 5 factors out of running this program, that are stored in ./etc/global gain scalefactors.dat
- The executable is called 'calibrator' and takes just one argument, the raw ROOT file with Am-241 data

Part#3: Use the two parts above, and generate a simple gainmatch file that is ready to use

- Basically, at this stage, we have all the S_{mn} values ready from part#1. We perform the chi2 minimization described in eq 5 and 6 in Reese(2015) now.
- Gives us a series of s_m values for all fronts, backs, all detector segments we've used in #1 up to a multiplicative constant. We can scale up each board's sn values by its own ' α ' by reading from the output txt file in #2 above
- All of this action is in src/fitter/fitter.cpp. Another cpp file which already has a main() in it, since the use-case is so simple. The executable is called 'findGainMatchFactors' which takes as argument the run# and the board#
- The shell script which wraps it, basically calls it for all 5 detectors/boards.
- The neat trick is that it reads outlist.0, outlist.1 ... outlist.4 as well as the global_gain_scalefactors.dat to then create s_n such that 1ch = 1keV holds for all 5 detectors.
- The final output is temporarily written to a textfile called "outputFile". But at the end of the analysis, it gets moved to the folder runXXX with all other data as gainmatch.runXXX – a simple text file with two columns, ready for use with GWM EventBuilder