README for Image Classification by Matt Bower, Armin Grossrieder

OUR ALGORITHMS

Naive Bayes

- for each image in the training set, extract features (pixels) and store in frequency table
 - for each test image, $P(D=d \mid E) = P(E \mid D=d) P(D=d)$

Perceptron

- for each image
 - for each pixel
- P(D=d | E) = (#samples(total) * #samples(XIY)) / (#samples(X) *
 #samples(Y))
- took this from a graphical perspective; saw there were x amount of input nodes
 and y amount of output nodes
- made predictor a matrix function; weights of connections between inputs and outputs * inputs matrix
- we took this matrix ____ and we made a cost function where we took the predicted Y values subtracted by the actual Y
 - then we made a cost y function to find the gradient of the function
- then we passed the objective function and the derivative to the pystat optimize function to give us the optimal set of weights that would give the best predictions Neural Network
 - saw that the perceptron performed much worse than Bayes formula
- sought out to improve it to add a hidden layer into the perceptron model in order to increase the accuracy of the model
- added a signmoid function in order to increase the accuracy- this would highlight the values that had the most effect on the output
- after doing this, it dramatically increased the run time, so we had to limit the amount of call backs the optimizer could make such that we wouldn't overload main memory
 - this led to impractical run times
- this model can only be used with a GPU to take advantage of the matrix multiplications

PROBLEMS

- probabilities not adding up
- probabilities equating 0 (before smoothing)
- running out of memory
- debugging... unsure whats going on
 - solution: struct and print_struct methods
- accuracy
- formatting data for algorithms
- everything just runs slower on the second iteration on any given loop...