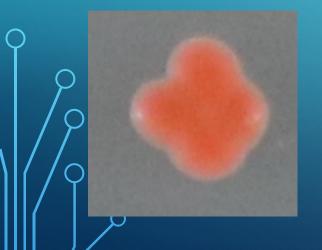
DEVELOPMENT OF A YEAST COLONY CLASSIFICATION SYSTEM USING COMPUTER VISION.

GREG GLICKERT

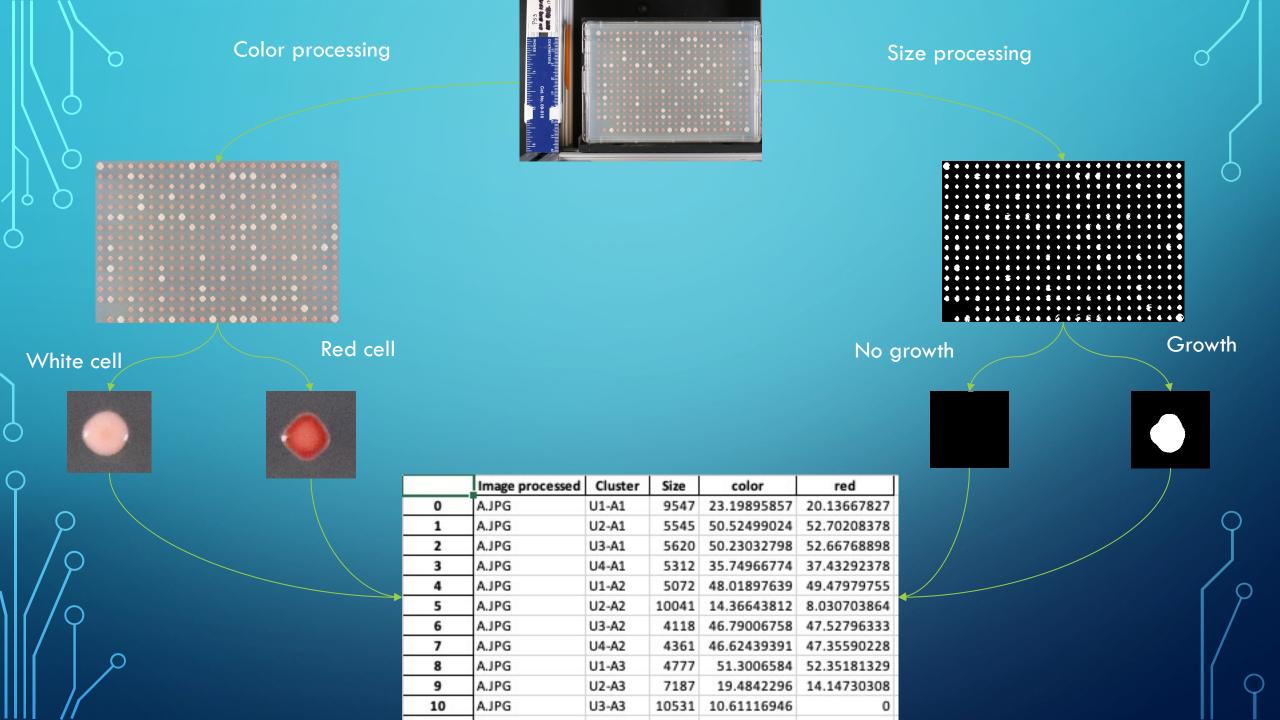
PROBLEM STATEMENT

High throughput phenotyping generates a large amount of data and classifying this data by hand presents many problems such as being tedious and bias.









TECHNICAL OVERVIEW

- Script is written in python 3.7
- Script runtime is around 80 seconds per image on a 2.3 GHz X86_64 core
 - Biggest contributor to the runtime is finding the plate which can be improved
- Dependencies for script are
 - Pillow, cv2, NumPy, os, pandas, matplotlib, imutils, SciPy, easygui, tqdm, xlrd, SKlearn

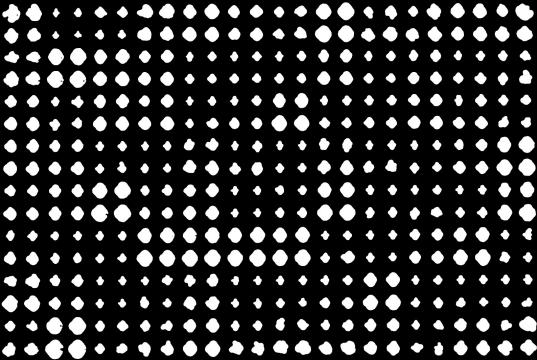
CROPPING THE PLATE

- 96 well, 384 or in the future 1536 well plates can be cropping into cluster of four or into individual cells
- User selects how they want the image cropped
- Uses the pillow library



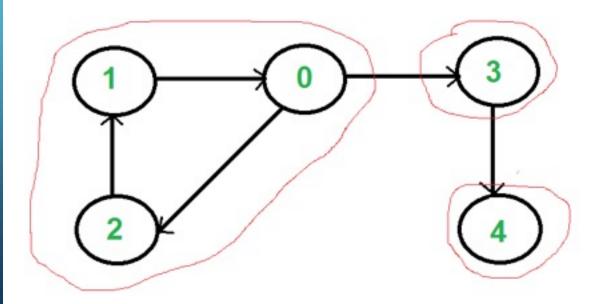
THRESHOLDING

- Takes the cropped image and turns background black and cells white
- Uses open computer vision library to threshold and remove anything that turns white that's not a cell



HOW THE SIZE IS FOUND

- The size is found using a connected components algorithm available in the open computer vision library
- Algorithm gives the size of each cell and the location of each cell



HOW COLOR DATA IS FOUND

- Color data is split into channels then analyzed to return the color metric
- Redness determined similar but first all color but red is removed from the image
 # split the image into its respective RGB components

(B, G, R) = cv2.split(image.astype("float")) # CV2 works in BGR not RGB # compute rg = R - G rg = np.absolute(R - G) # compute yb = 0.5 * (R + G) - B yb = np.absolute(0.5 * (R + G) - B) # compute the mean and standard deviation of both `rg` and `yb` (rbMean, rbStd) = (np.mean(rg), np.std(rg)) (ybMean, ybStd) = (np.mean(yb), np.std(yb)) # combine the mean and standard deviations stdRoot = np.sqrt((rbStd ** 2) + (ybStd ** 2)) meanRoot = np.sqrt((rbMean ** 2) + (ybMean ** 2)) # derive the "colorfulness" metric and return it return stdRoot + (0.3 * meanRoot)

WHAT HAPPENS TO THE DATA?

- The data is sent to a pandas data frame
 - Can be exported into xlsx or csv format

	2	Image processed	Cluster	Size	color	red
	0	A.JPG	U1-A1	9547	23.19895857	20.13667827
	1	A.JPG	U2-A1	5545	50.52499024	52.70208378
)	2	A.JPG	U3-A1	5620	50.23032798	52.66768898
	3	A.JPG	U4-A1	5312	35.74966774	37.43292378
	4	A.JPG	U1-A2	5072	48.01897639	49.47979755
	5	A.JPG	U2-A2	10041	14.36643812	8.030703864
	6	A.JPG	U3-A2	4118	46.79006758	47.52796333
	7	A.JPG	U4-A2	4361	46.62439391	47.35590228
/	8	A.JPG	U1-A3	4777	51.3006584	52.35181329
	9	A.JPG	U2-A3	7187	19.4842296	14.14730308
	10	A.JPG	U3-A3	10531	10.61116946	0

	Image processed	Cluster	Q1_size	Q2_size	Q3_size	Q4_size	Avg_size
0	01.JPG	U1-A1	2938	3807	4959	4792	4124
1	01.JPG	U1-A2	2129	2965	3777	4080	3237.75
2	01.JPG	U1-A3	3484	3903	4407	4653	4111.75
3	01.JPG	U1-A4	2656	3089	3343	3285	3093.25
4	01.JPG	U1-A5	3090	3563	3728	3727	3527
5	01.JPG	U1-A6	2686	3334	3664	3753	3359.25
6	01.JPG	U1-A7	2808	3252	3527	3765	3338
7	01.JPG	U1-A8	2772	3115	3446	3715	3262
8	01.JPG	U1-A9	1407	1706	1979	2208	1825

WHAT CAN YOU DO WITH THIS DATA?

- Train a machine learning model to learn and predict which cells are red and which cells are white.
- Improvements
 - More labeled data
 - Feature engineering
 - More advanced models

Confusion Matrix for svc [[15 0] [4 135]] accuracy for svc 0.974025974025974

	Predicted O	Predicted 1		
Actual O	TN	FP		
Actual 1	FN	TP		

Confusion Matrix for mlp [[15 0] [4 135]] accuracy for mlp 0.974025974025974