

# COSC 499 Milestone 4

Team 9 order of aesthetics - section 003

Algorithm used to compare  
images

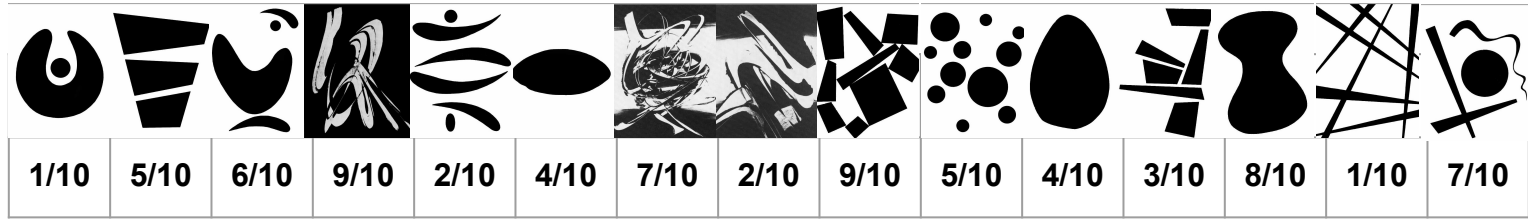
# Recall: How to build a neural network

Start with unlabeled data

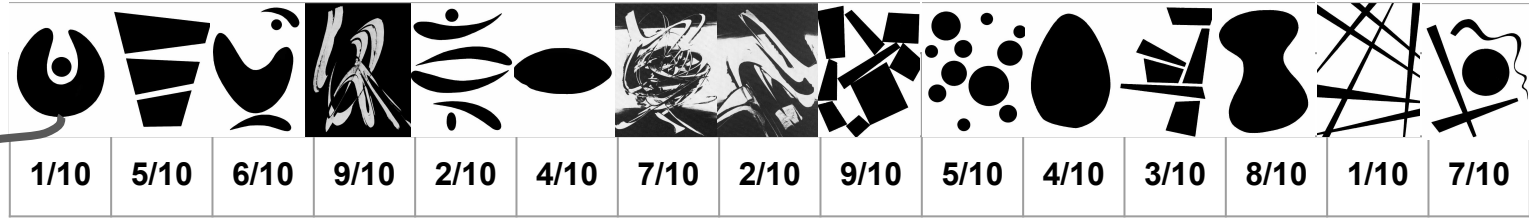


# Recall: How to build a neural network

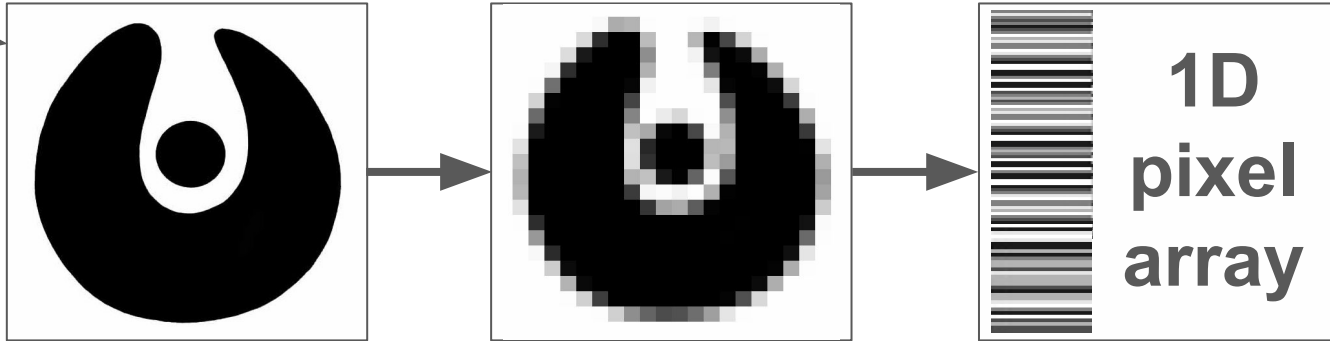
Label it using the survey in our apps



# Recall: How to build a neural network

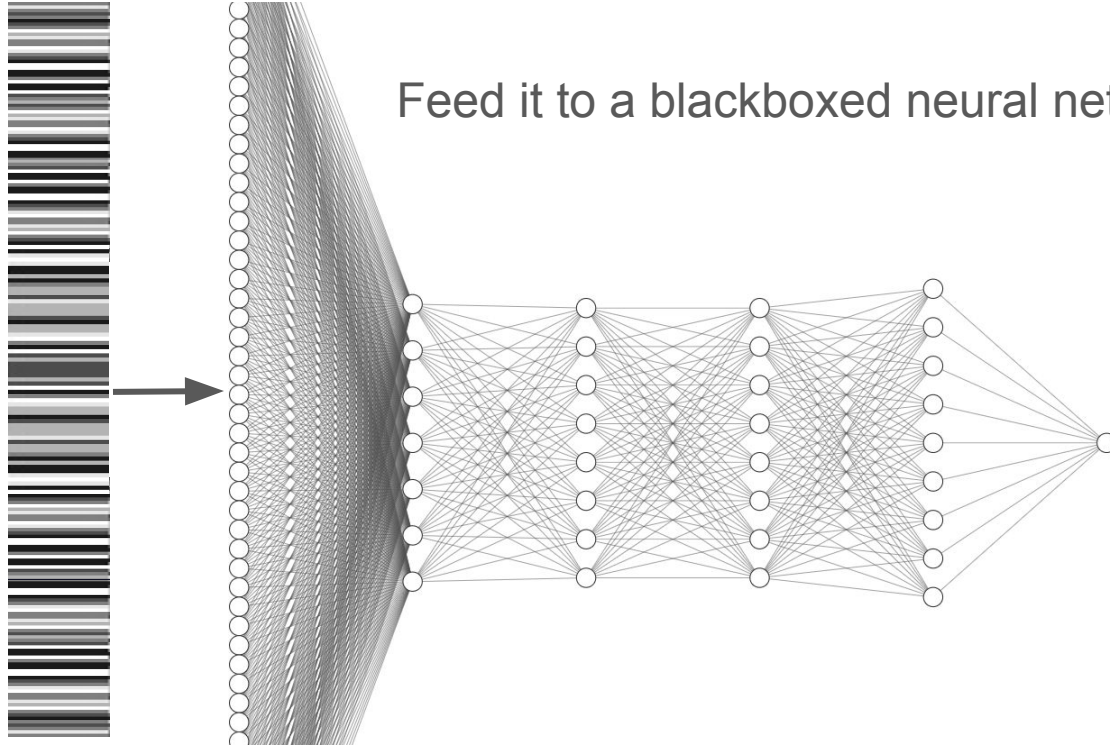


Convert each image to a set of pixels (pretraining)



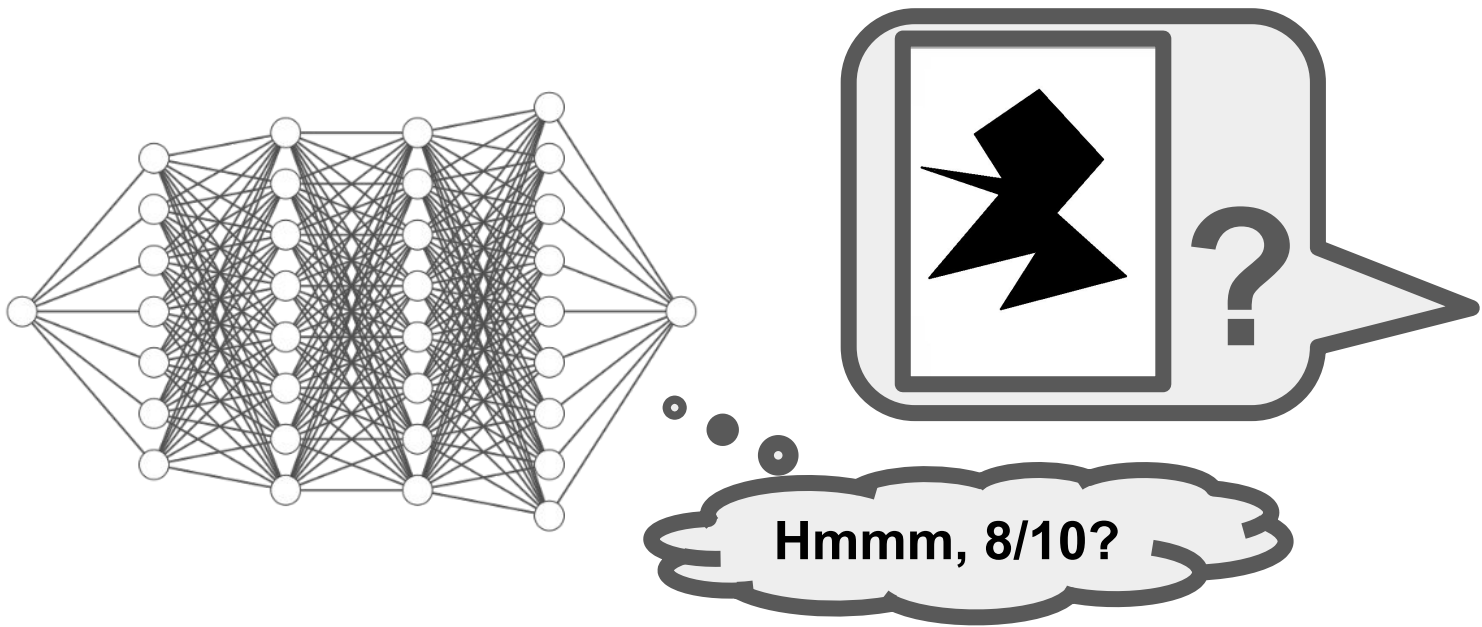
# Recall: How to build a neural network

**1D  
pixel  
array**



# Recall: How a neural network works

And just like that we have a neural network that can make an educated guess as to how aesthetically pleasing an image is



# Performance of image comparison



How do we measure performance?

How do we measure performance?

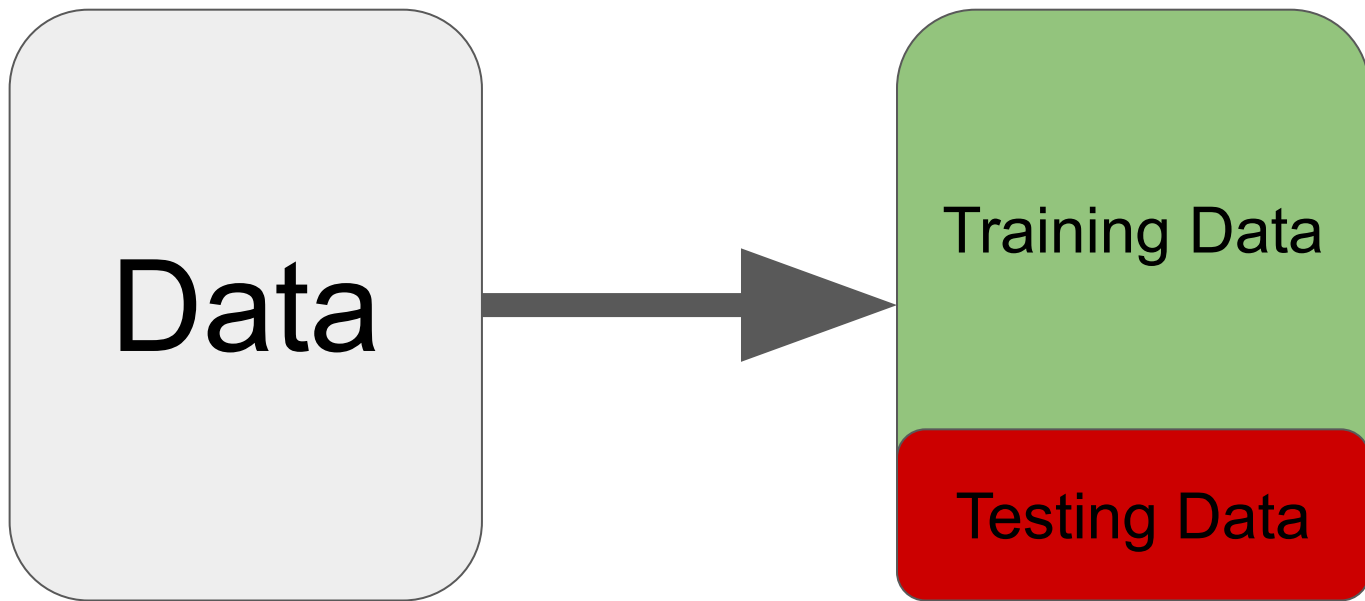
We implement a Train/Test split in our data

Train test split:

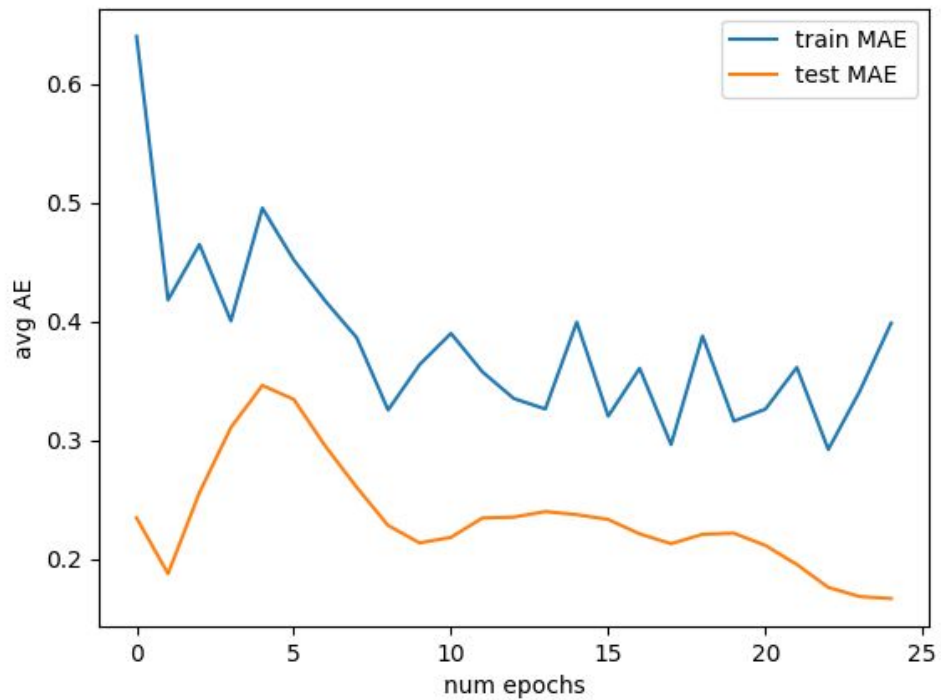


**Data**

Train test split:



Train test split:



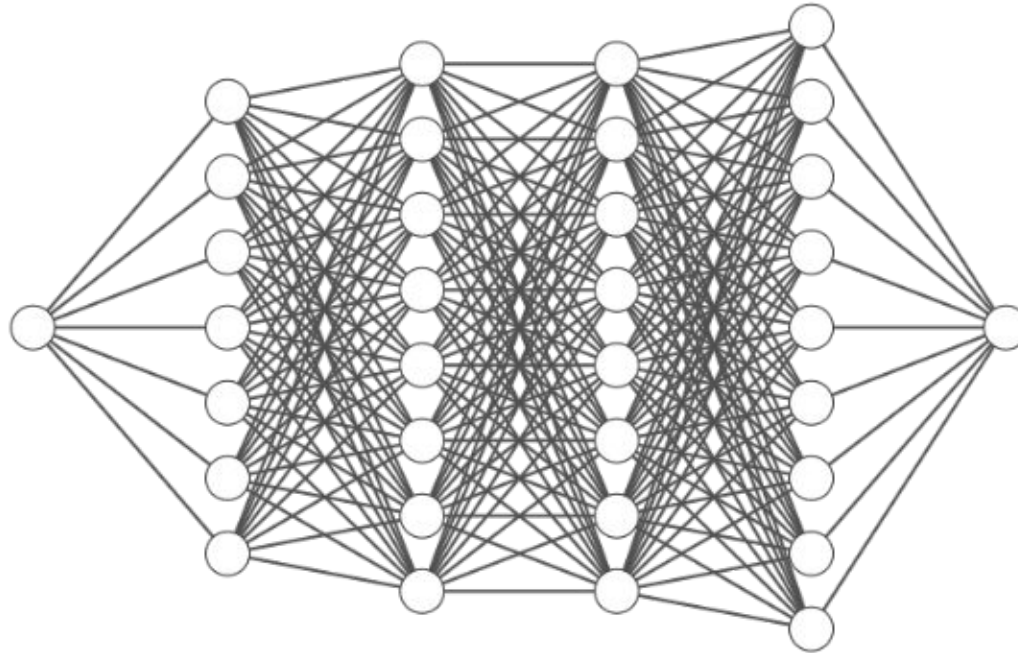
# So Ultimately:

Our algorithm had an average error of 20% while trying to guess which image out of a pair was preferred by the majority of users (while guessing would be 50% +/-)

Meaning: for every 10 images, the Algorithm would align with the majority opinion for 8 of them (while guessing would have gotten 5 of them)

Algorithm used to generate  
images

# Starting out with our algorithm from earlier



Input Layer  $\in \mathbb{R}^1$

Hidden Layer  $\in \mathbb{R}^7$

Hidden Layer  $\in \mathbb{R}^7$

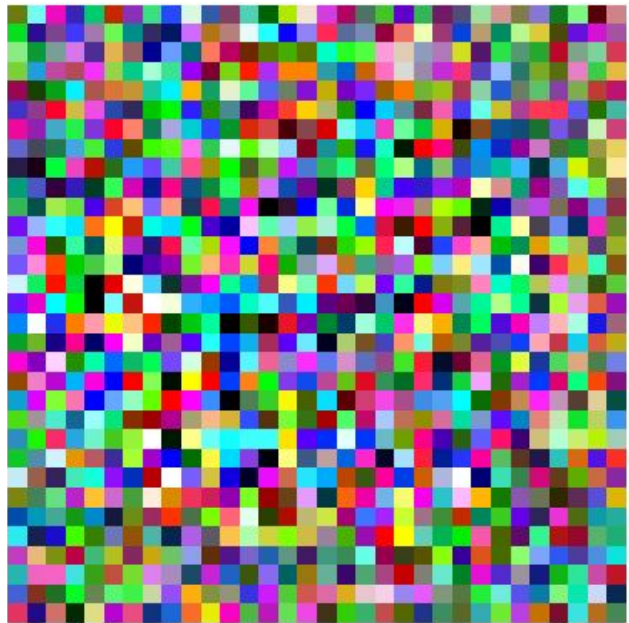
Hidden Layer  $\in \mathbb{R}^7$

Hidden Layer  $\in \mathbb{R}^7$

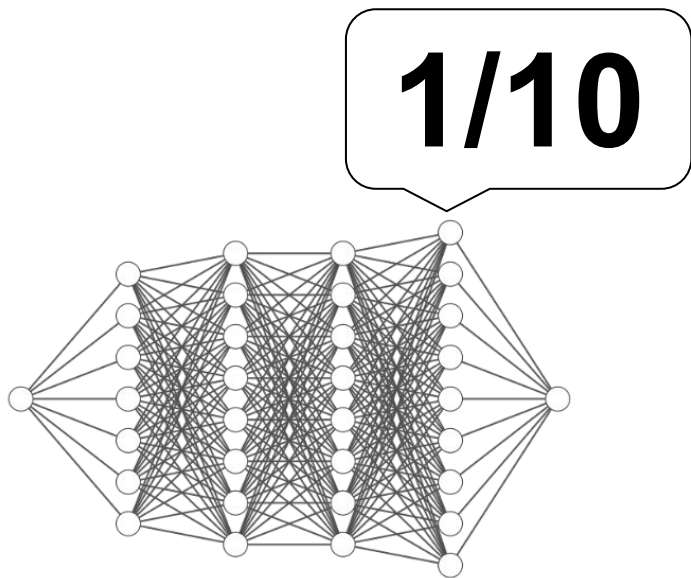
Output Layer  $\in \mathbb{R}^1$



We generate an image from statistical noise

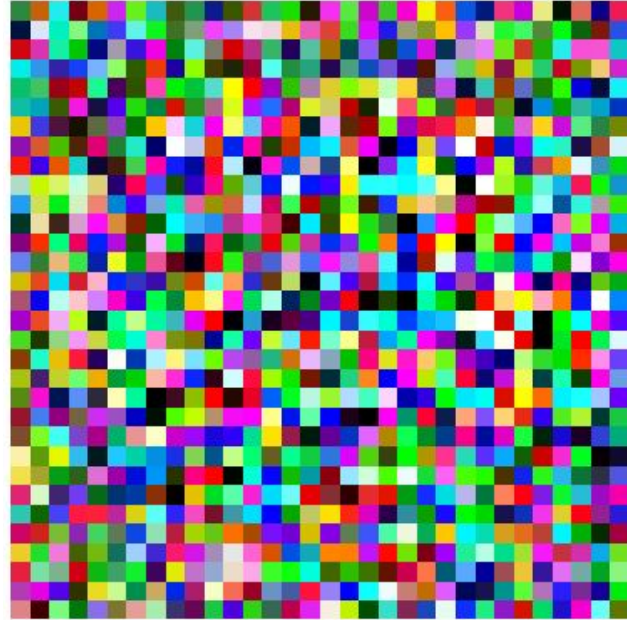
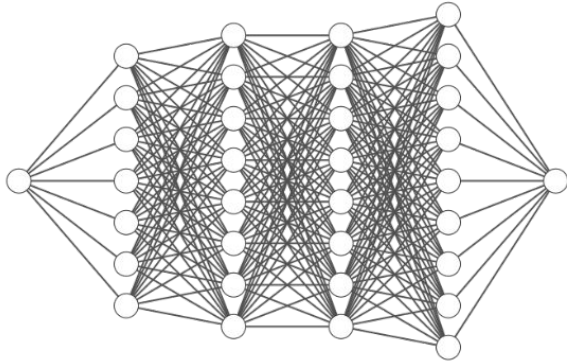


& ask our model what it thinks

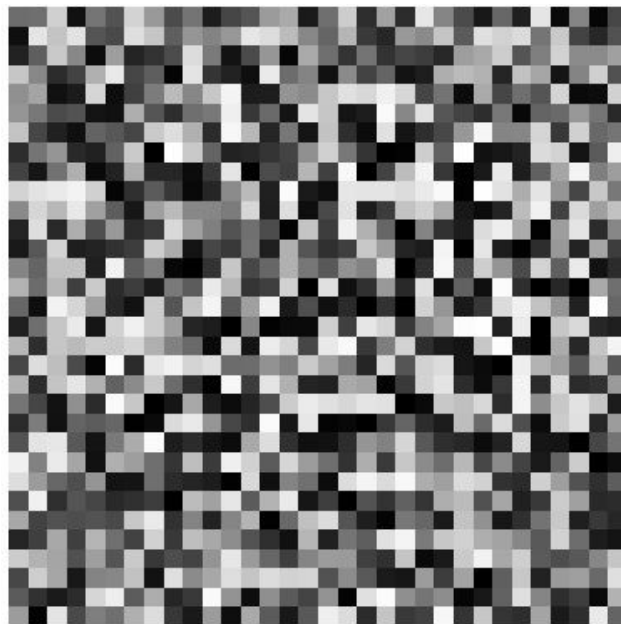
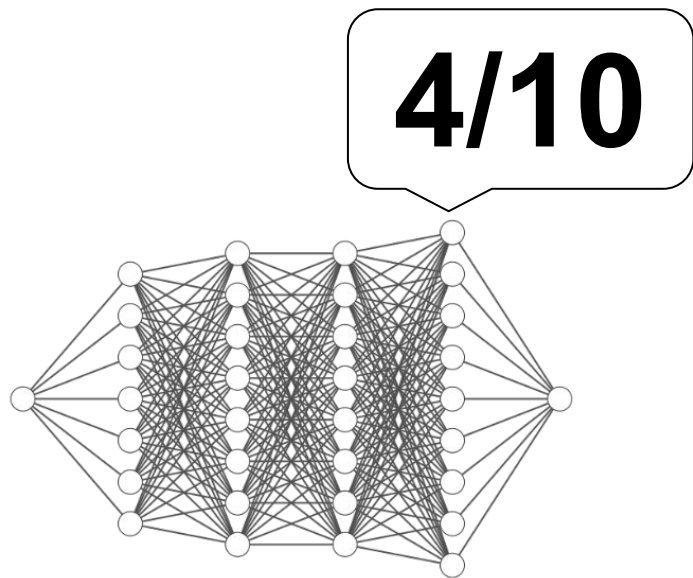


Using that we can generate a better image

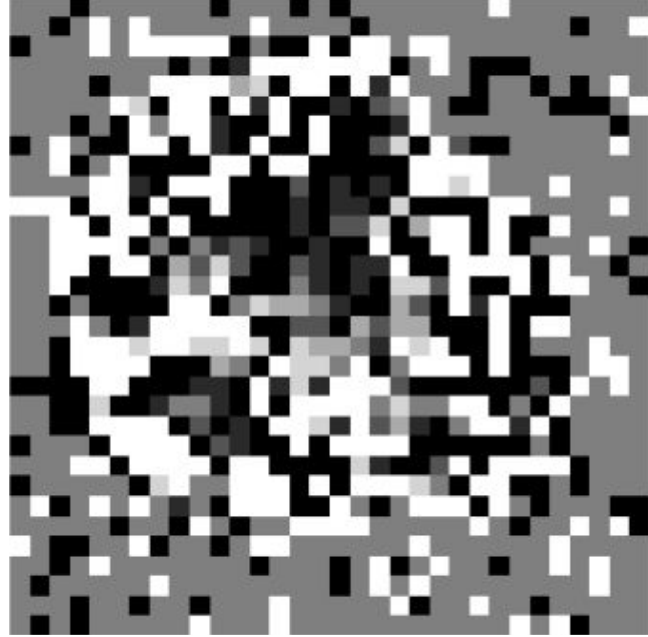
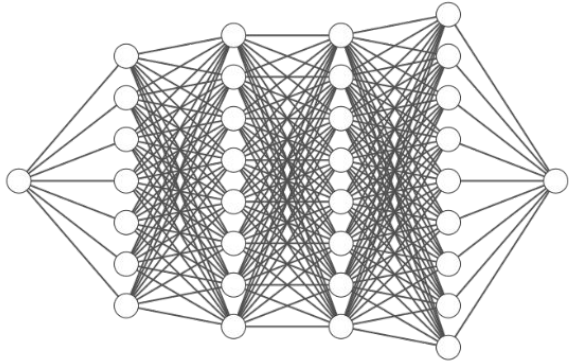
**2/10**



& then an even better image

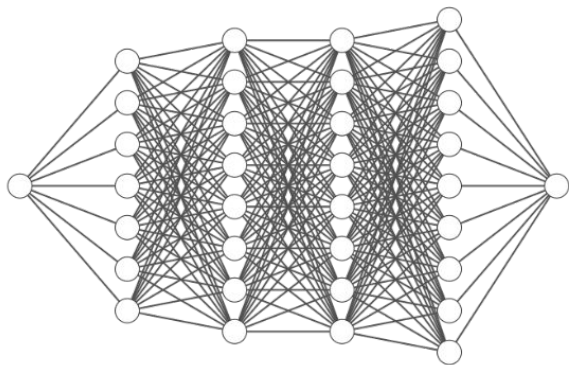


**7/10**

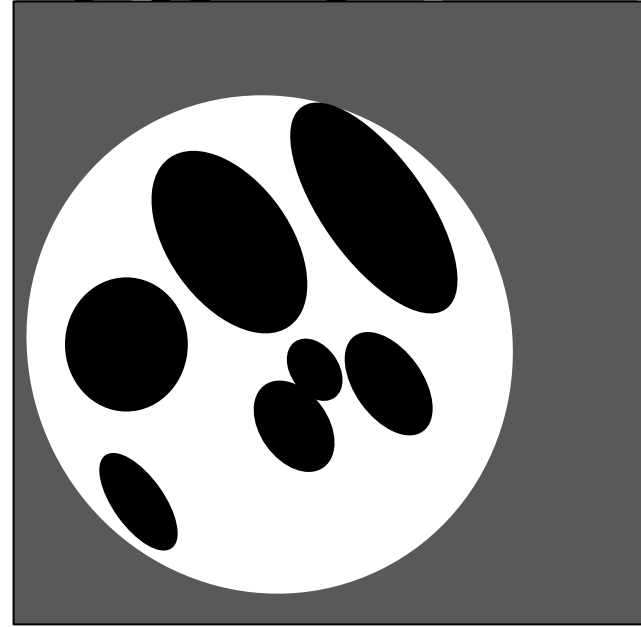


& then an even better image

**8/10**



Then finally we can smooth to create a final product

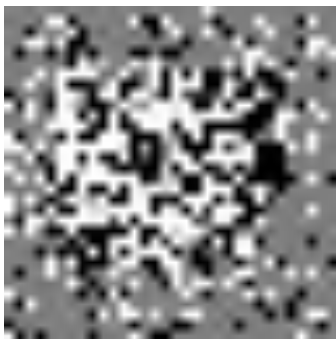
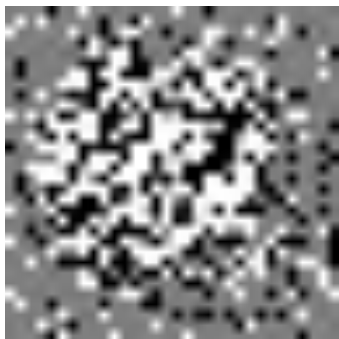


Smoothing aims to take a set of raw math output,  
& interprets it into an image:

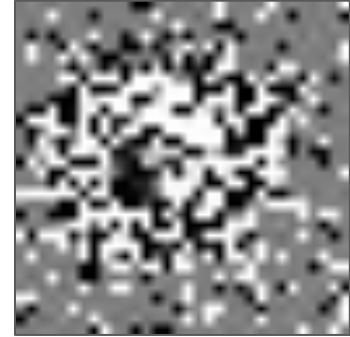
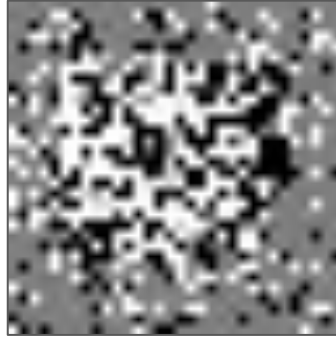
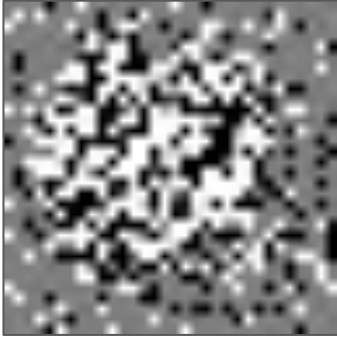


Smoothing aims to take a set of raw math output,  
& interprets it into an image:

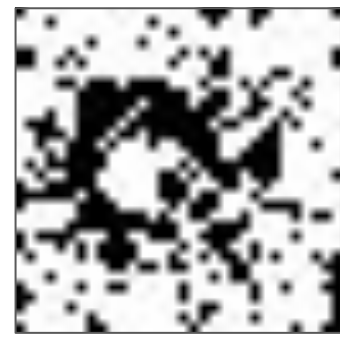
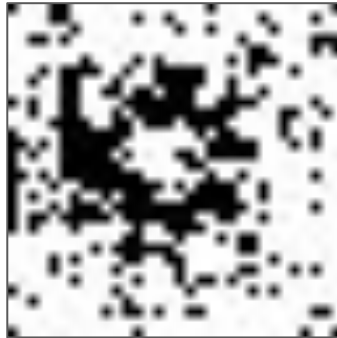
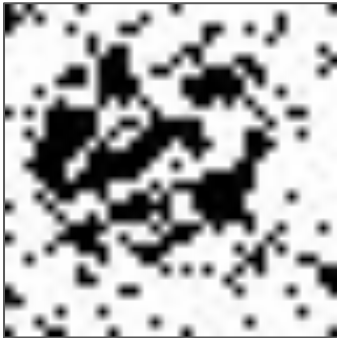
Our generated images ended up looking like this raw:



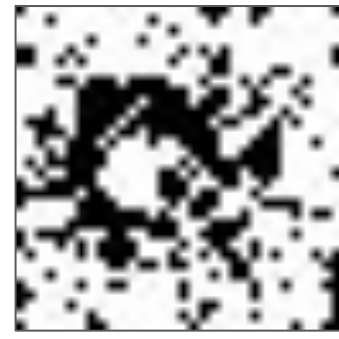
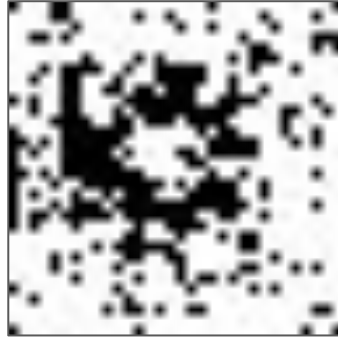
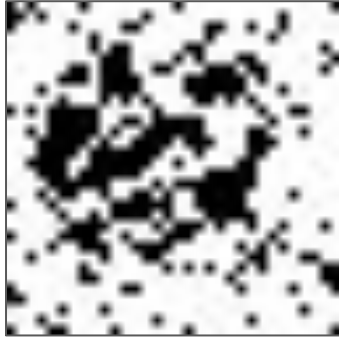
Our generated images ended up looking like this row:



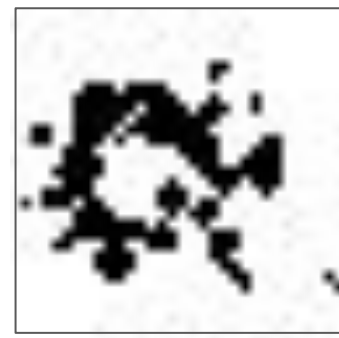
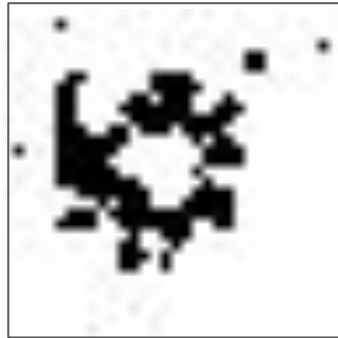
And like this after applying 4 Nearest Neighbor smoothing once:



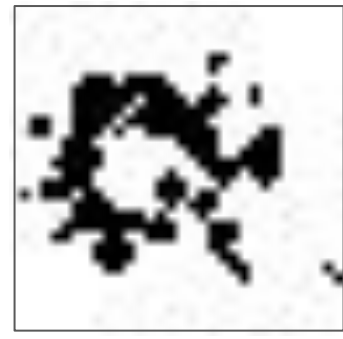
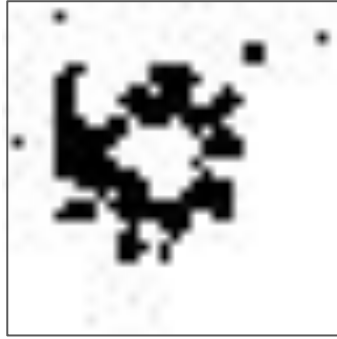
And like this after applying 4 Nearest Neighbor smoothing once:



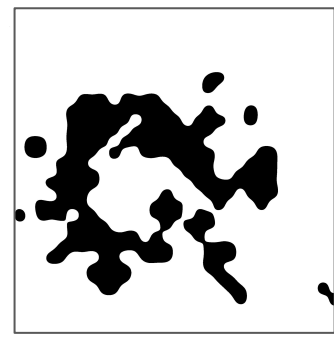
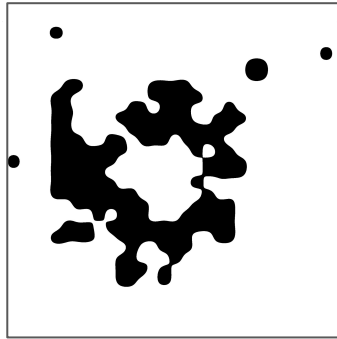
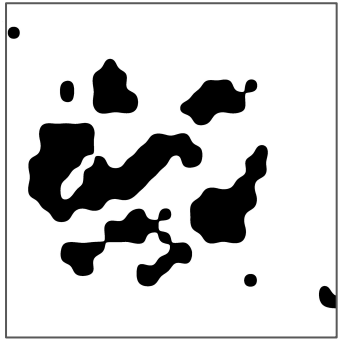
Then like this after applying Smoothing again:



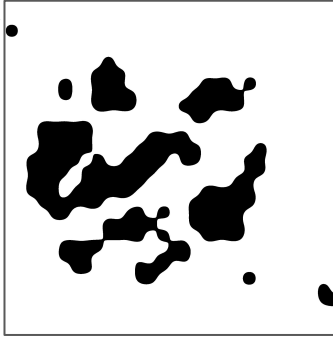
Then like this after applying Smoothing again:



Then like this after using interpolation to make HD versions:

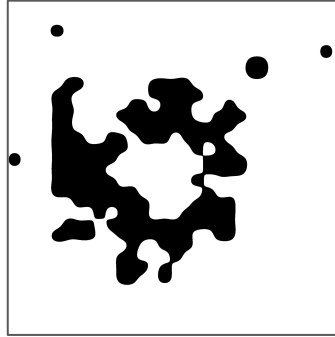


And rated by 5 people on a scale from 1-10:



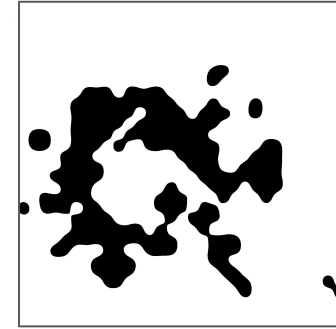
$$(9+8+10+7+9)/5$$

$$= 8.6/10$$



$$(7+8+9+8+7)/5$$

$$= 7.8/10$$



$$(6+8+9+5+8)/5$$

$$= 7.2/10$$

Therefore: the algorithms in place can make adequate but non excellent images - this serves as a proof of concept that with a sufficient pool of data a successful algorithm can generate images en mass

# A note on small sample size

# Thanks for listening

## A note on dovetailing:

Some elements in this slideshow are copied from team member Samira Almuallim's work in the client information session 3 & the M3 slides