



NEXT PROJECT

WRITING A THESIS ...

„EXPLAINABLE AI FOR STEM CELL MODELS WITH LEIGH DISEASE“

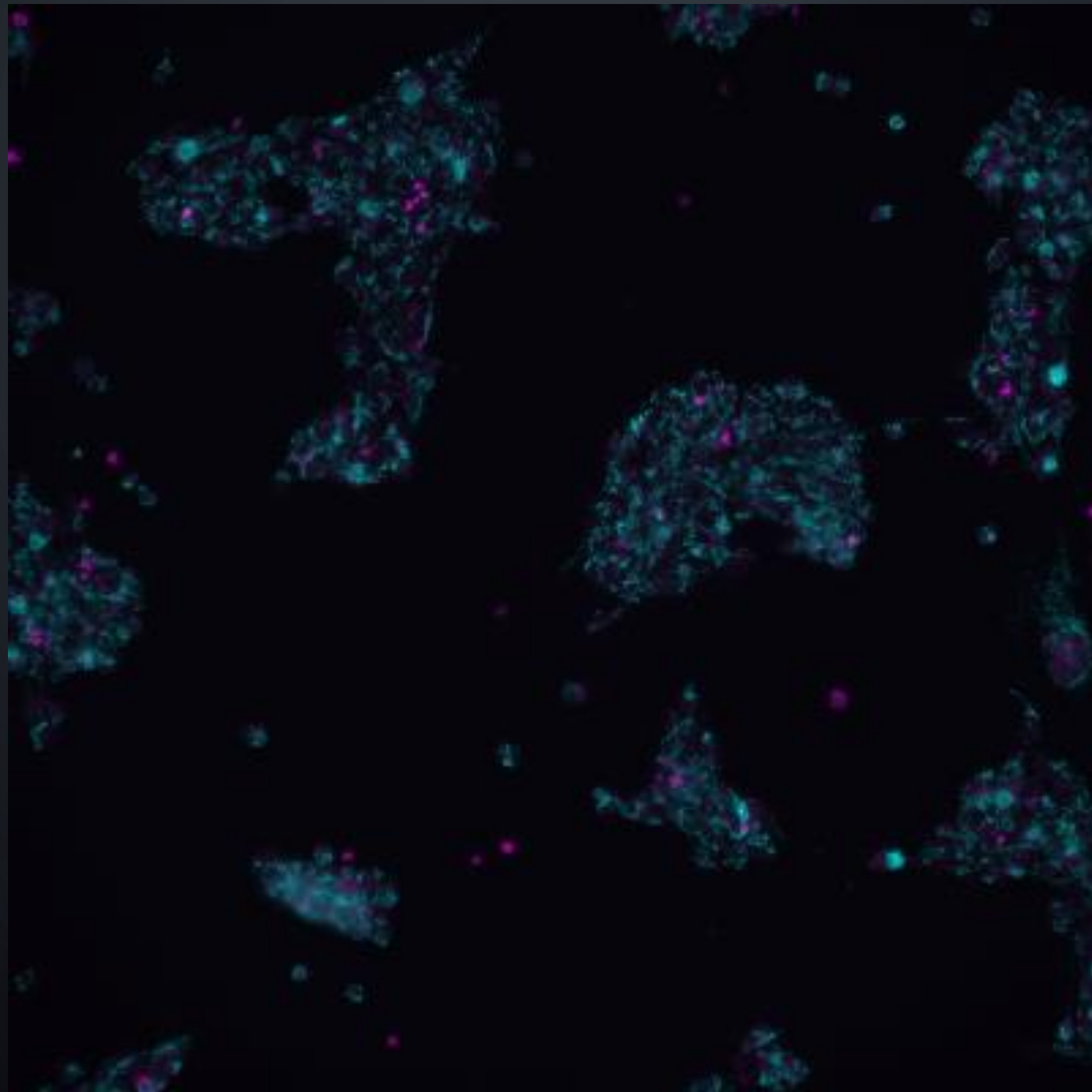


Previous Project:

Classifying Leigh Disease Images



Combined image:



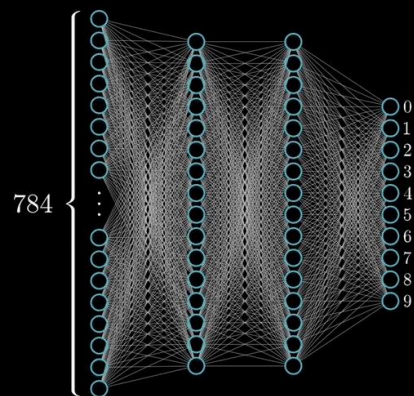
CLASSIFICATION (DS VS. TT)

Got 540/540 with accuracy 100.000%

	precision	recall	f1-score	support
DS	1.00	1.00	1.00	264
TT	1.00	1.00	1.00	276
accuracy			1.00	540

Z'-Factor: 0.924

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WHAT ARE WE GOING TO DO WITH THIS?

- > Help to find the best component/drug for treating Leigh Disease
 - > scan for cell culture images appearing to be wildtypes after treatment
- > Try to understand the difference between the classes

HEAT-MAPS CAN BE SUFFICIENT, SOMETIMES



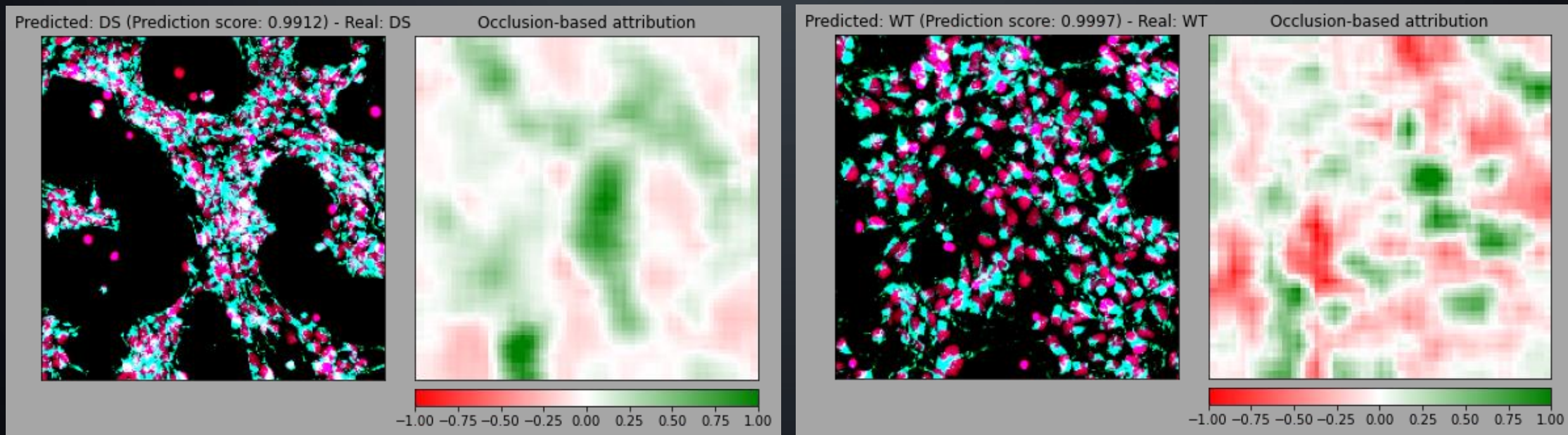
Figure 2: Example of a saliency map explanation of a True Positive (TP) image for the label “train”. It highlights the contours of the lines below the train. A possible interpretation is that the CNN has learned to recognise trains when rails are present.



Figure 3: Example of a saliency map explanation of a False Positive (FP) image for the label “train”. A possible interpretation is that edges in the lower part appeared similar to rails, which could explain this error.

OCCLUSION BASED ATTRIBUTIONS

Sanity check: It can detect our cells



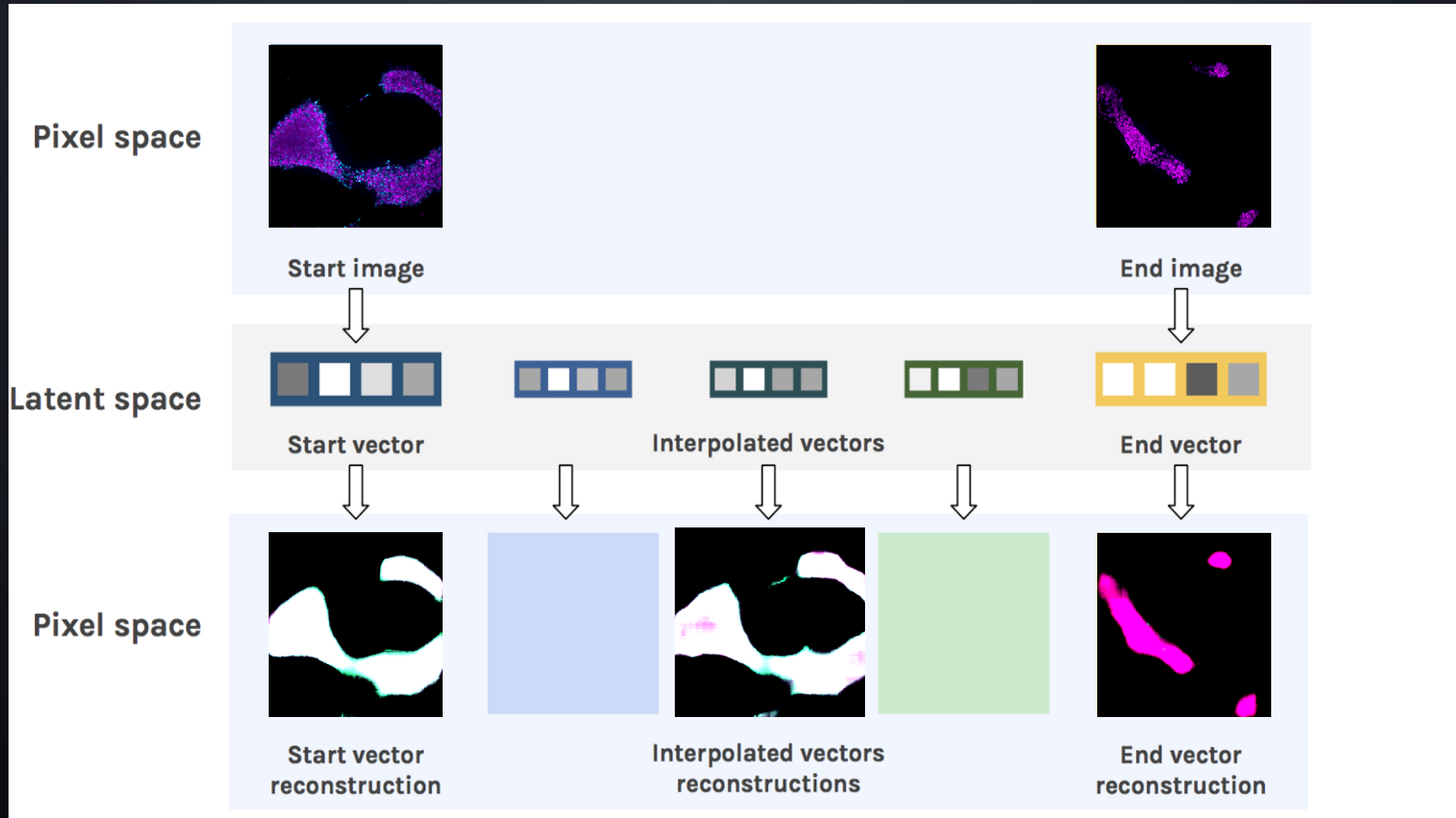
Problematic: In some images, empty areas are more important



NEW PLAN

- > Set up a neural network that can classify & generate images
- > Modulate an image to contain the information of another class
- > How does it change? Can we spot a difference?

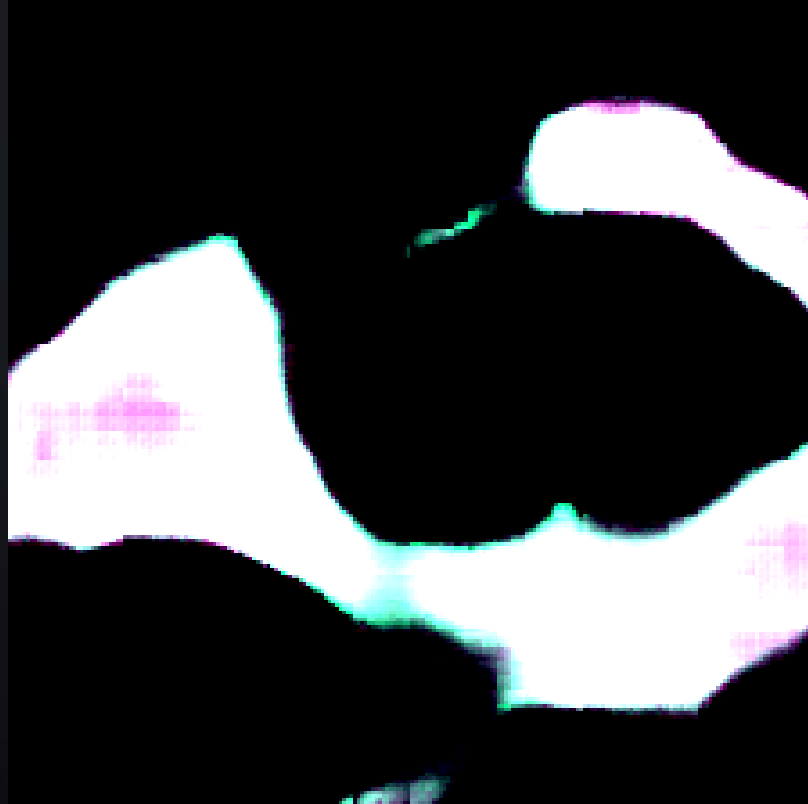
GENERATE NEW IMAGES (AUTOENCODER)



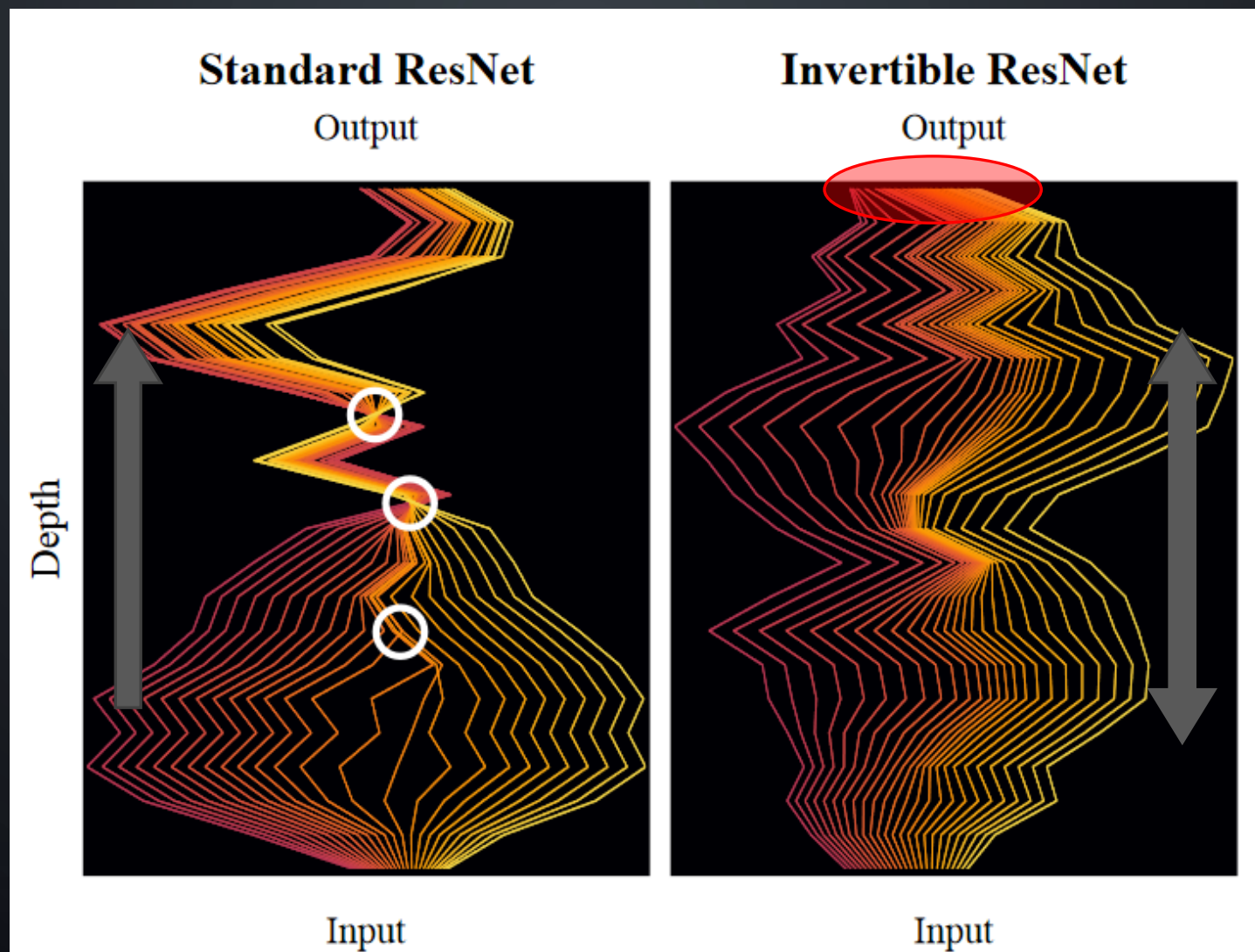
GENERATE NEW IMAGES (AUTOENCODER)



GENERATE NEW IMAGES (AUTOENCODER)



INVERTIBLE NEURAL NETWORK

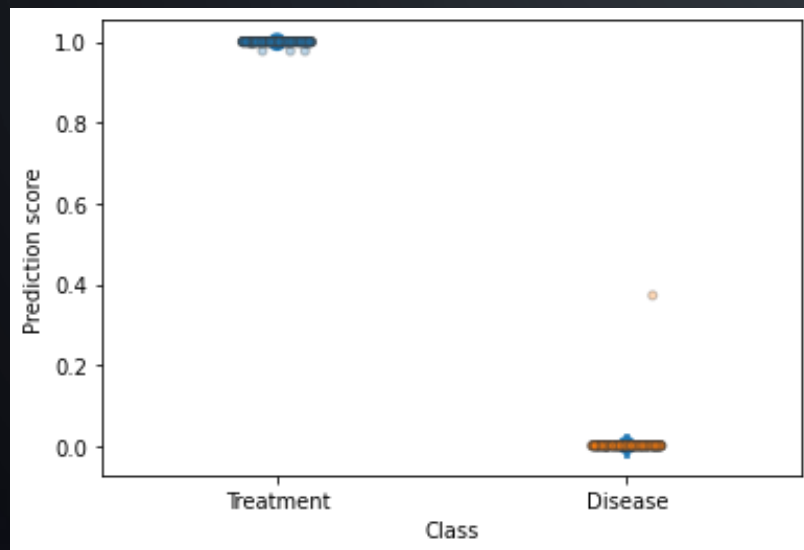


CLASSIFICATION (DS VS. TT)

ResNext50:

Got 540/540 with accuracy 100.000%

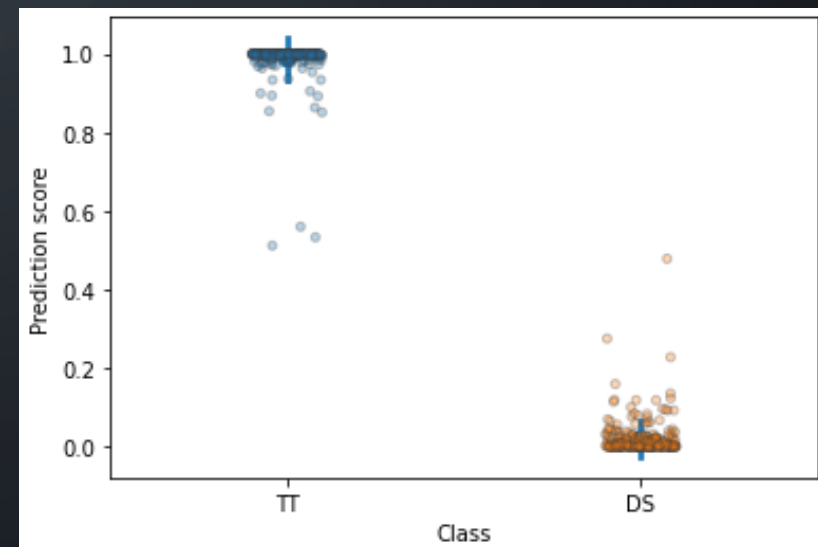
Z'-Factor: 0.924



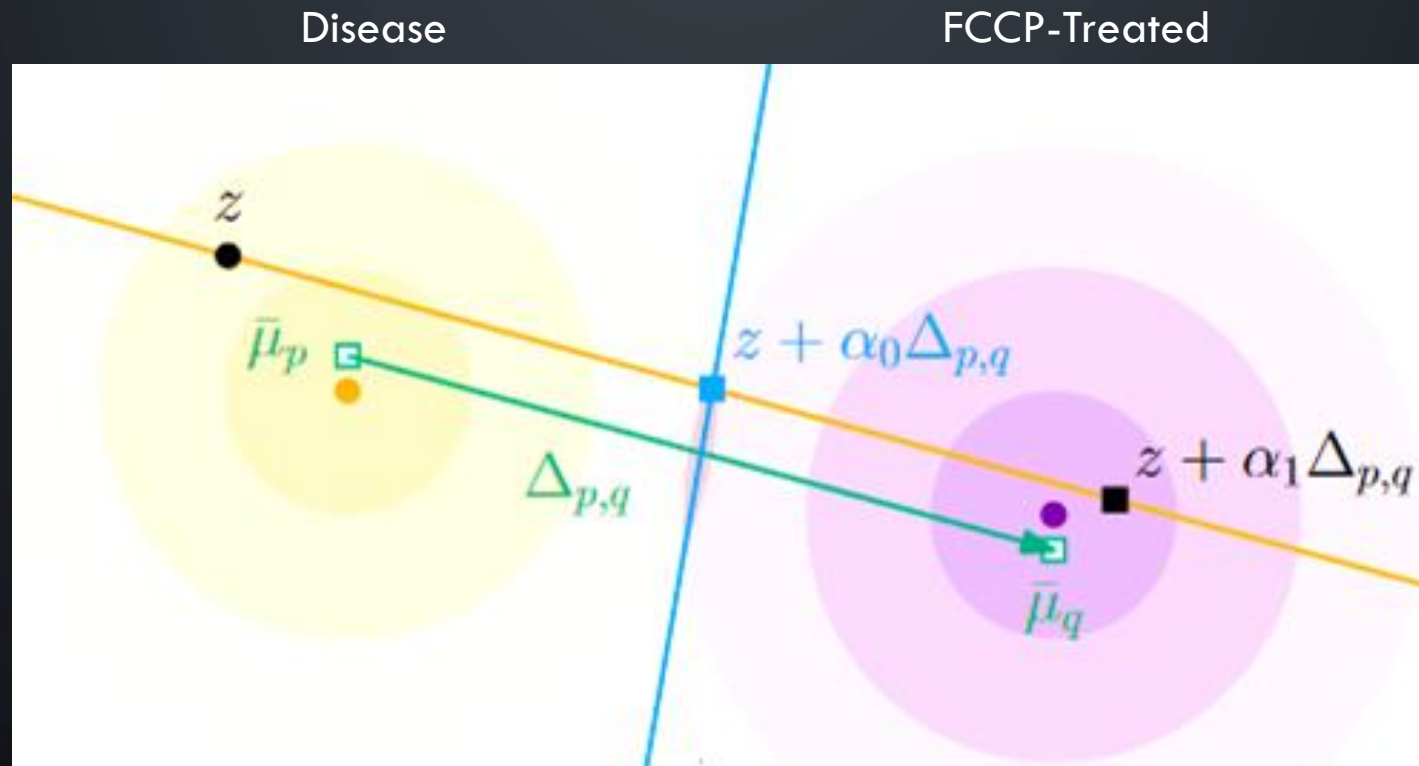
i-ResNet50:

Got 540/540 with accuracy 100.000%

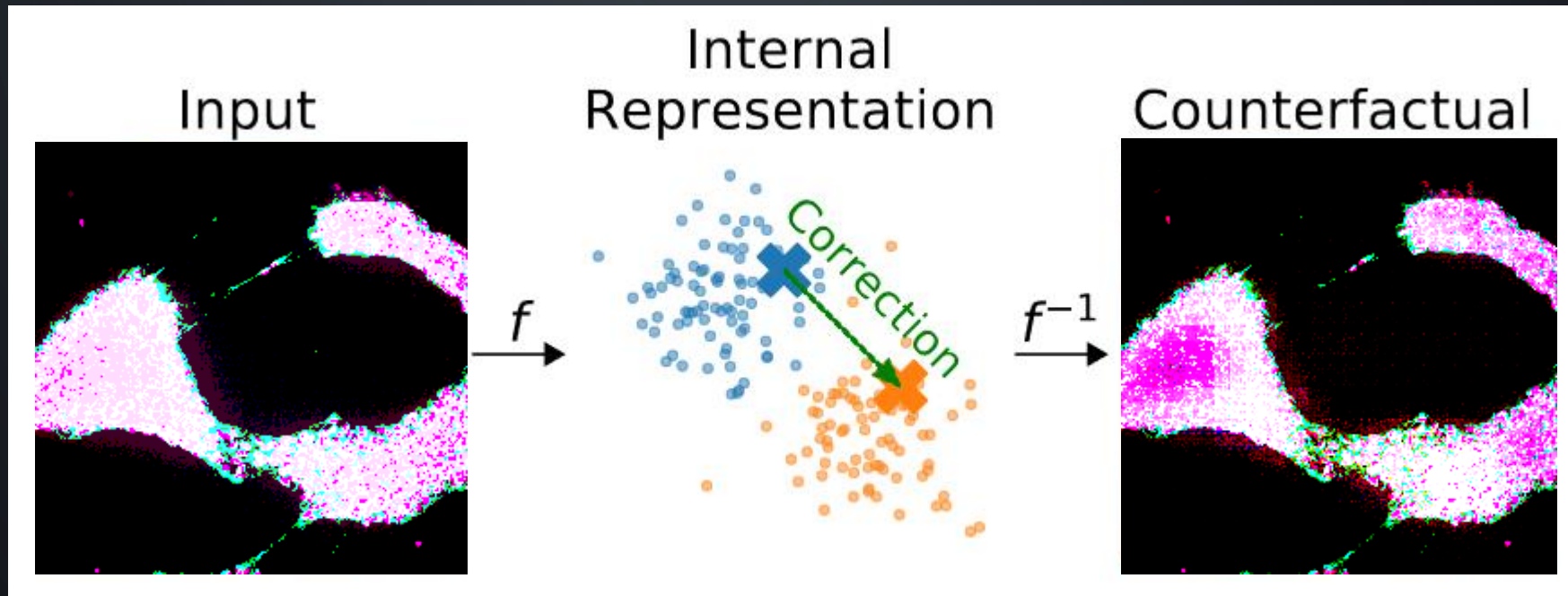
Z'-Factor: 0.701



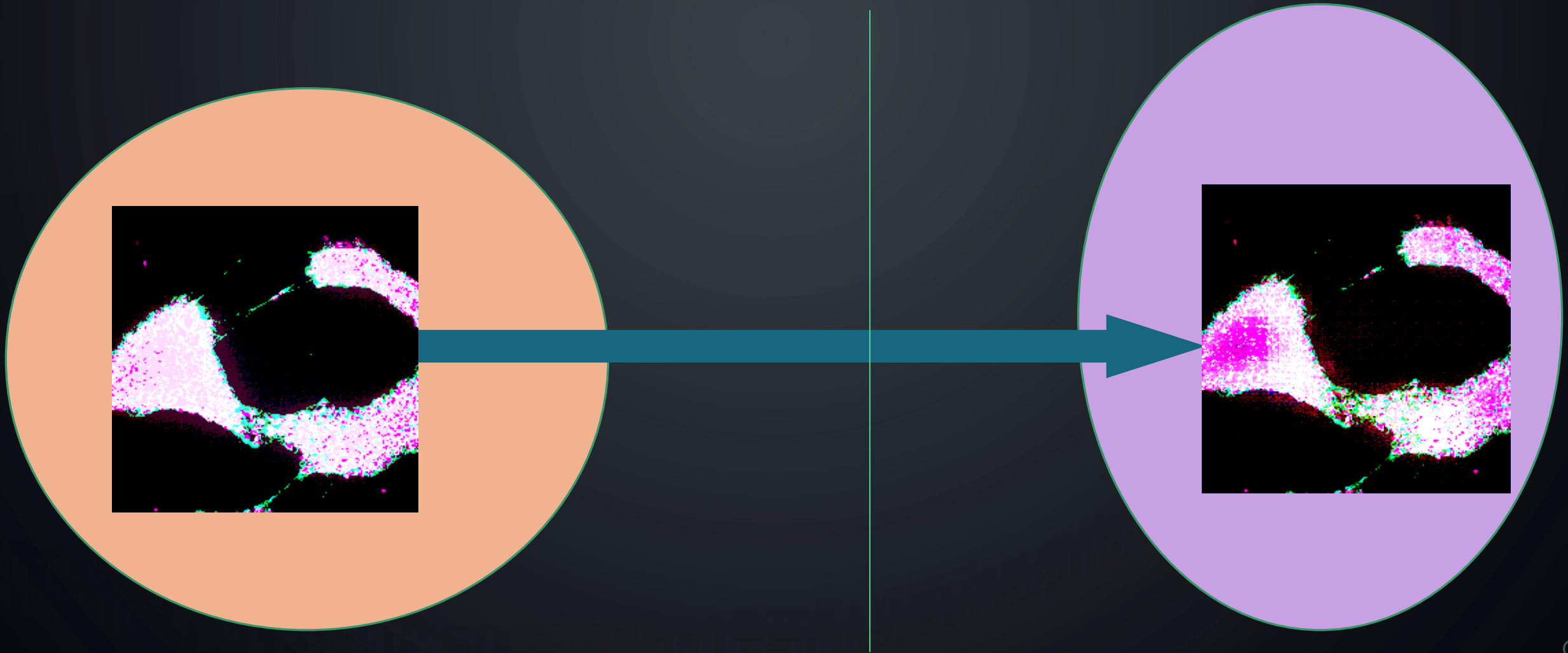
COUNTERFACTUALS



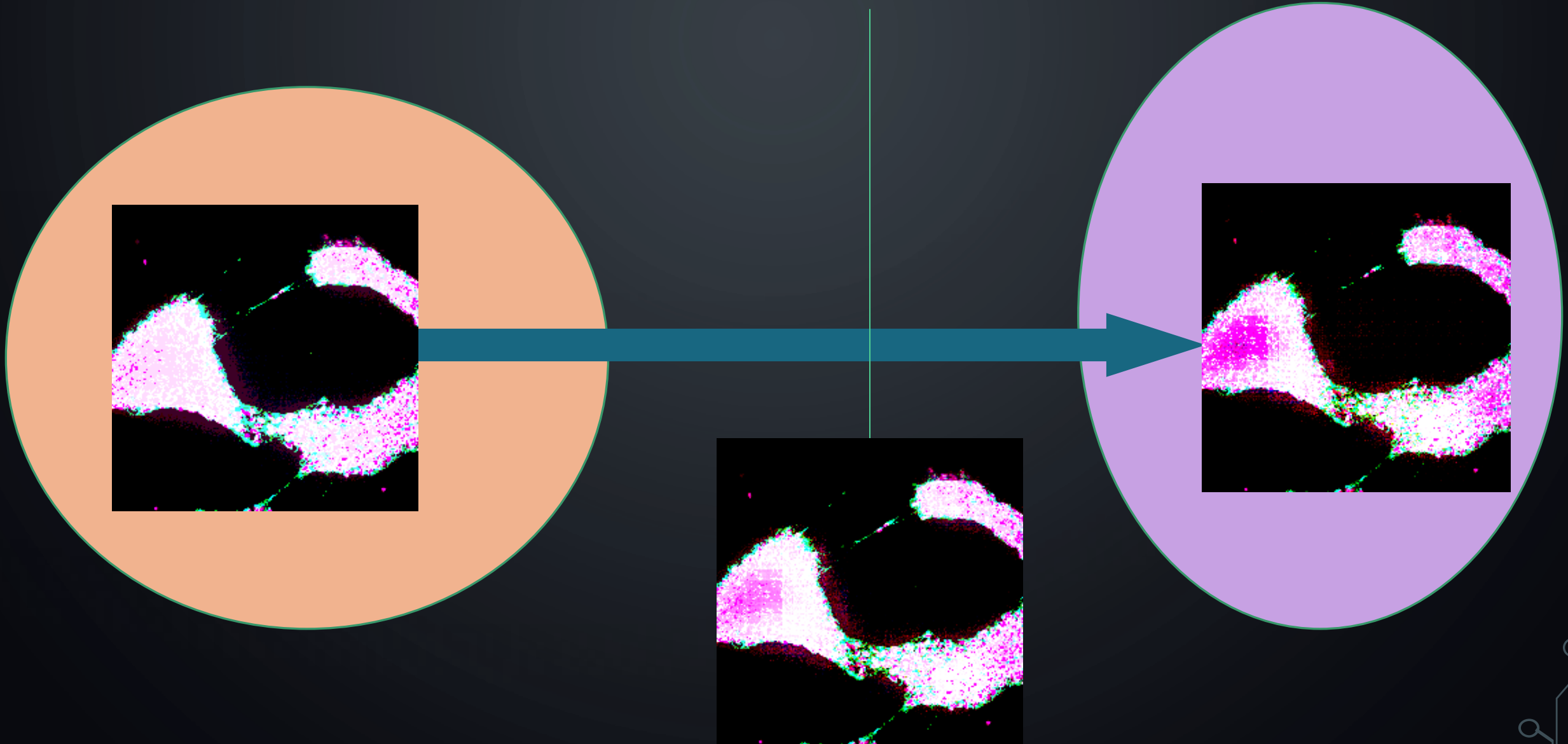
COUNTERFACTUALS



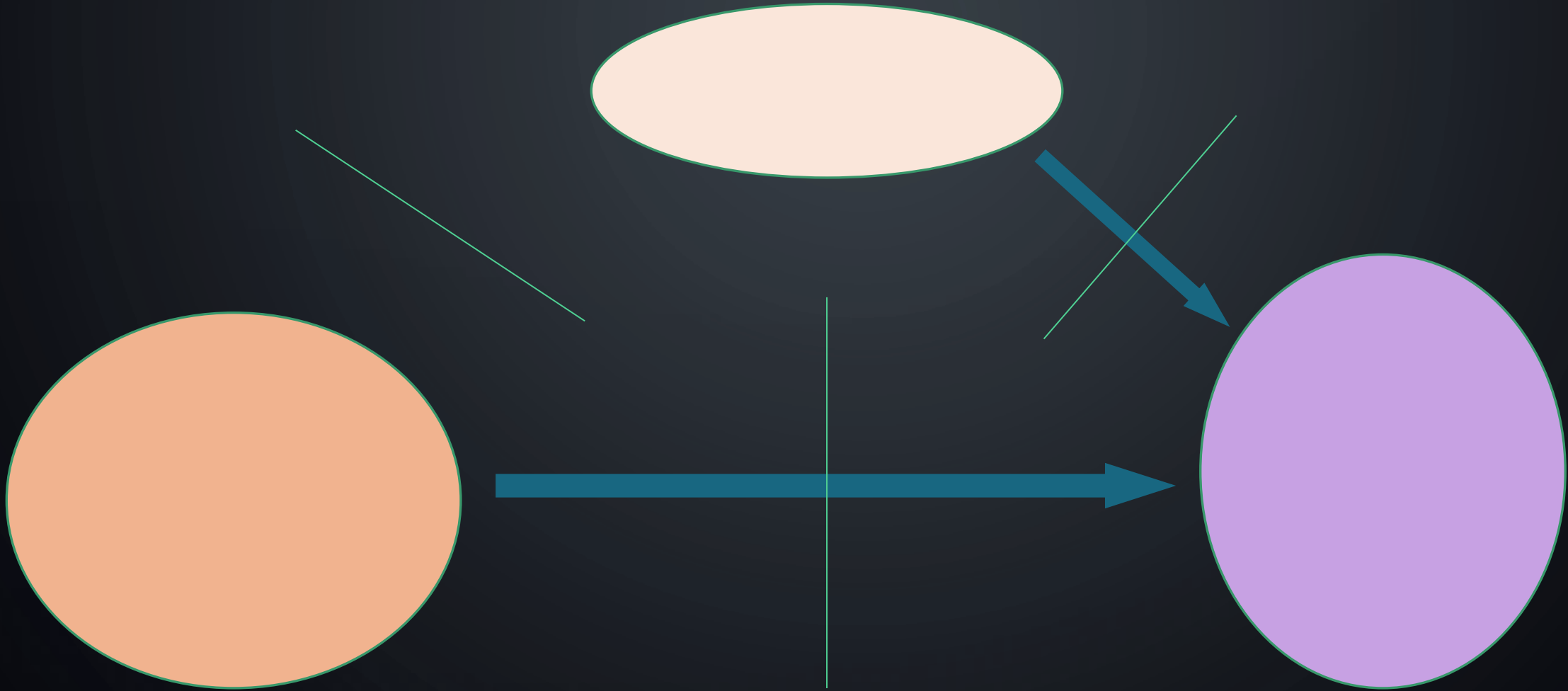
DECISION BOUNDARY




DECISION BOUNDARY



FUTURE WORK: MULTI-CLASS APPLICATION (DIFFERENT TREATMENTS)





> Set up a neural network that can classify & generate images

→ Found one

> Modulate an image to contain the information of another class

→ Convincing images

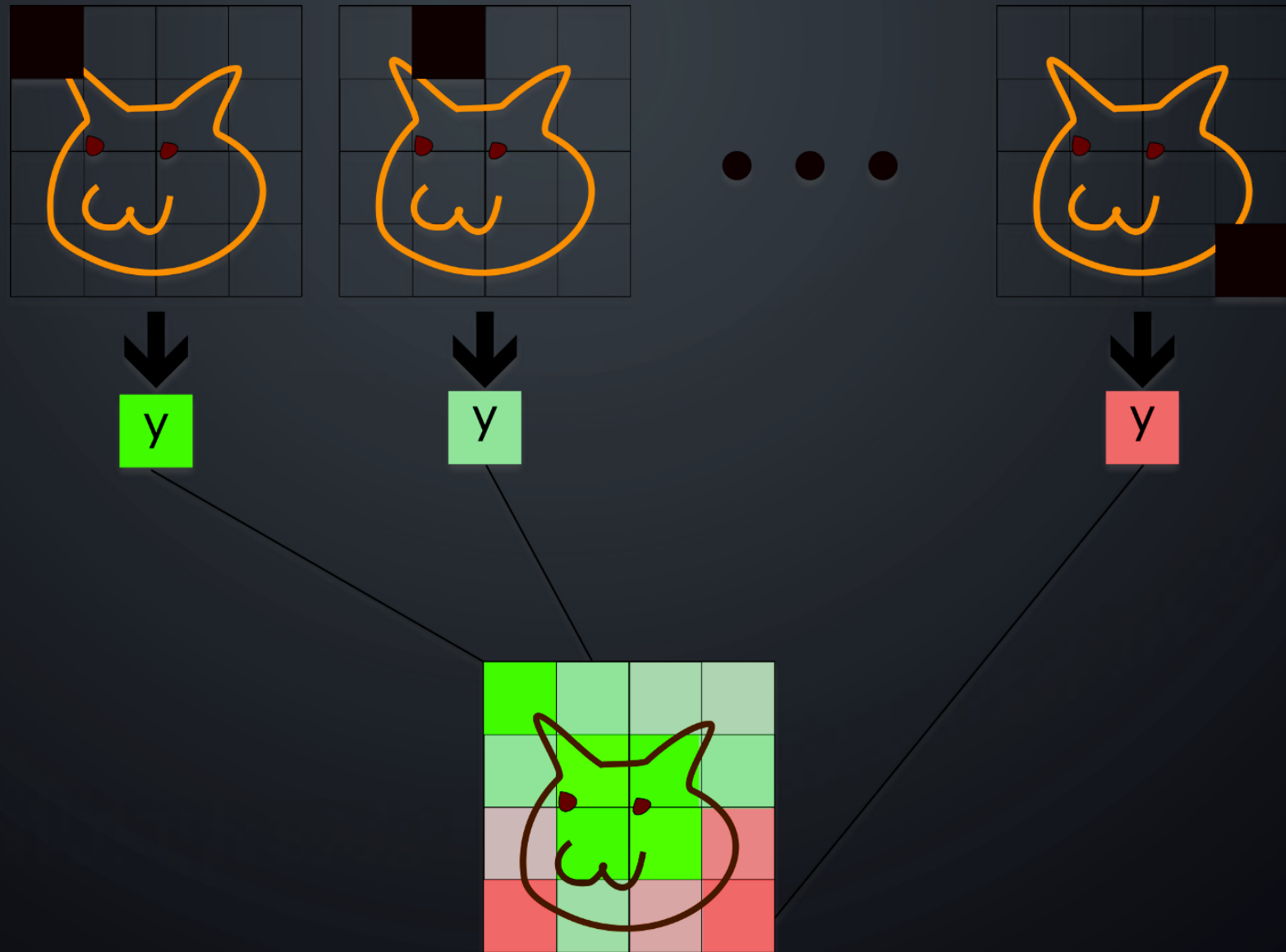
> How does it change? Can we spot a difference?

→ The treated cell cultures have a higher mitochondrial membrane activity (red)

→ Application for DS/WT is still ongoing

ATTRIBUTIONS

Occlusion Example:



AUTOENCODER

