

# Milky Way Challenge

Active Training Course „Advanced Deep Learning“

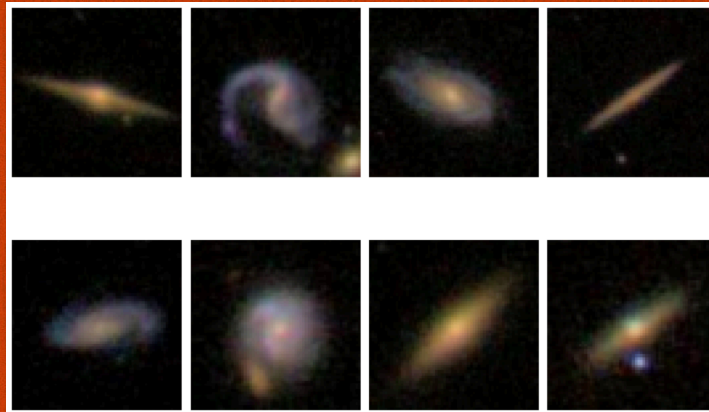
Meinerzhagen, December 1st, 2022

## Adversarial Attackers

Alex, Nathan, Sebastian, Stephan, Waleed

# Challenge

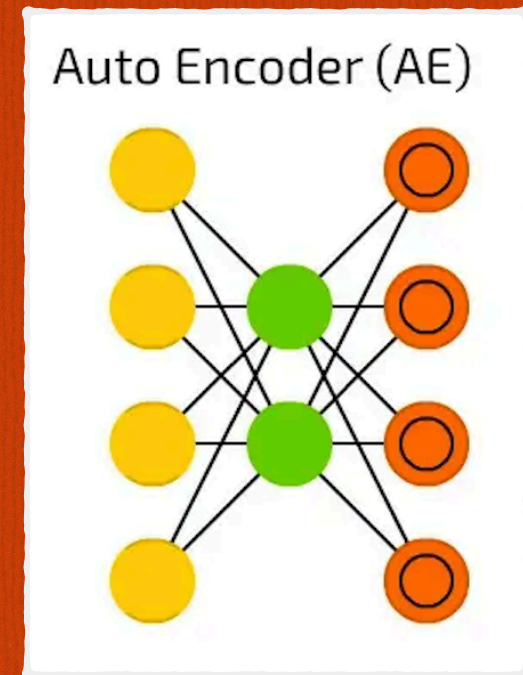
- **Dataset:** subset of „Galaxy10“ dataset (<https://astronn.readthedocs.io/en/latest/galaxy10sdss.html>)
- **Some modifications:** 4 classes removed, images cropped to 64x64 pixels



- **Task:** Find a galaxy that looks as close as possible to our Milky Way
- **Bonus task:** Generate new galaxies that look as close as possible to our Milky Way

# Main Idea

- Use a (variational) autoencoder
  - Encoder embeds features into latent space
  - Our idea: Features can be compared in latent space using euclidian distance
  - Decoder decodes features into pictures
- 
- Input image: 64x64 version of given Milky Way image in Google collab

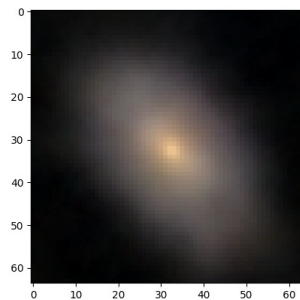
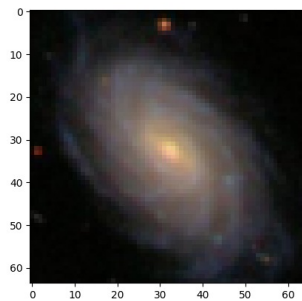


<https://towardsdatascience.com/the-mostly-complete-chart-of-neural-networks-explained-3fb6f2367464>

# Basic Idea: Autoencoder

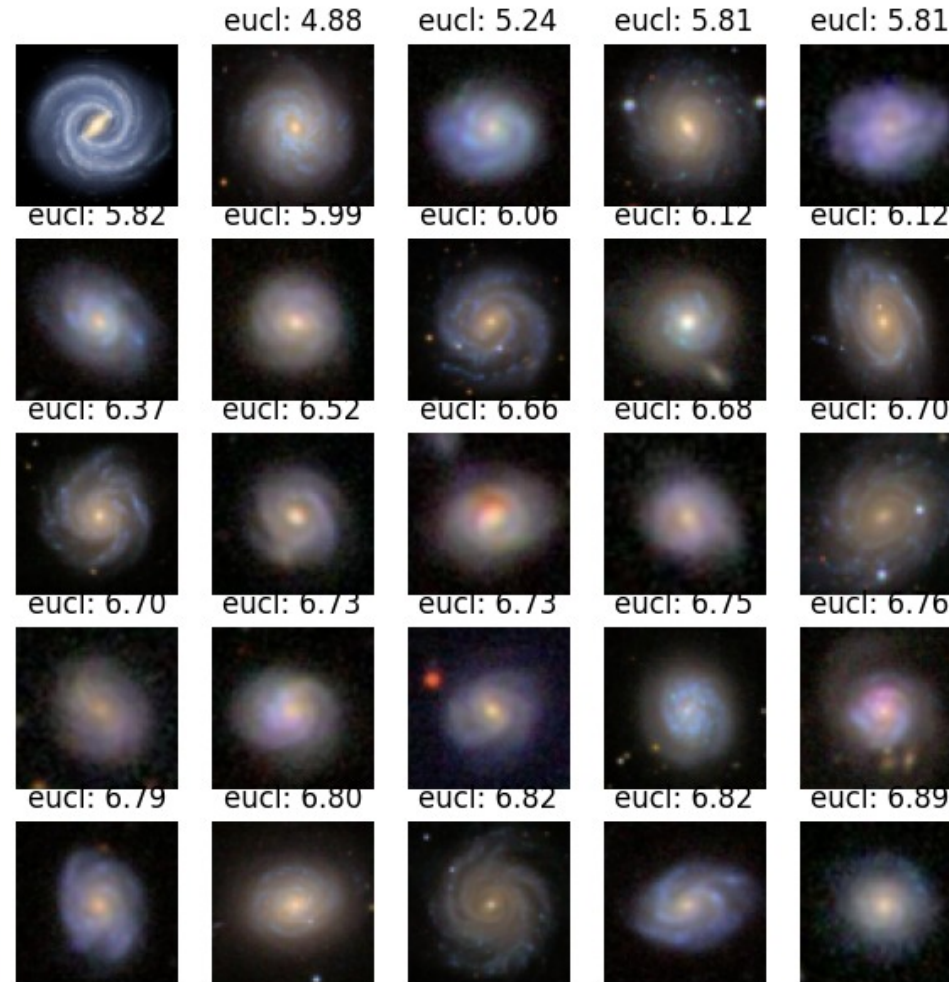
```
def build_autoencoder(img_shape, code_size):  
    encoder = Sequential(name = 'encoder')  
    encoder.add(InputLayer(img_shape))  
    encoder.add(Flatten())  
    encoder.add(Dense(code_size))  
  
    decoder = Sequential(name = 'decoder')  
    decoder.add(InputLayer((code_size,)))  
    decoder.add(Dense(np.prod(img_shape))) # hier 64 x 64 x 3  
    decoder.add(Reshape(img_shape))  
  
    return encoder, decoder
```

img\_shape = (64, 64, 3)  
Latent space dimension = 64

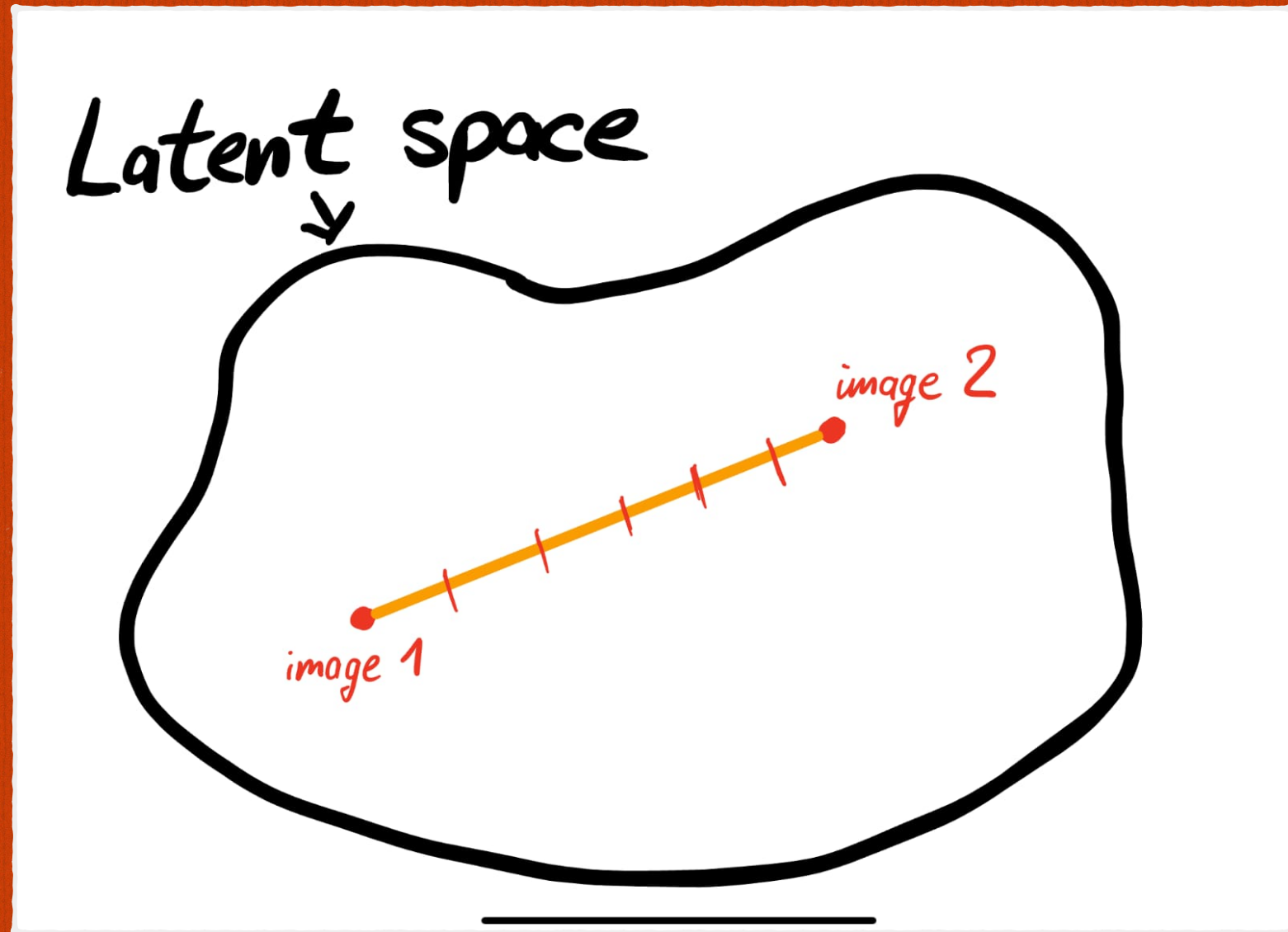


**Here could be your latent space**

# Autoencoder - similarity search



# Autoencoder - sampling idea

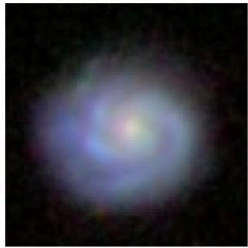


# Autoencoder - sampling

Interpolation: 0.00



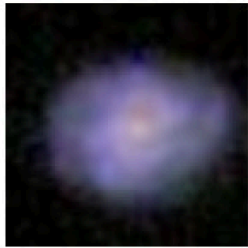
Interpolation: 0.20



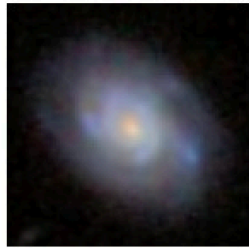
Interpolation: 0.40



Interpolation: 0.60



Interpolation: 0.80



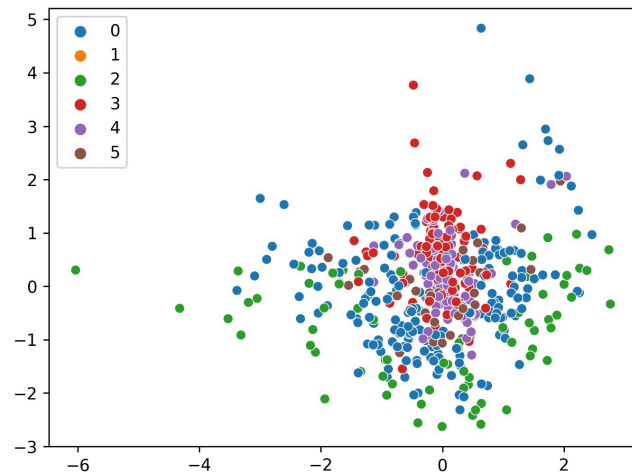
Interpolation: 1.00



# Variational autoencoder

```
ConvVAE(  
  (enc1): Conv2d(3, 8, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1))  
  (enc2): Conv2d(8, 16, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1))  
  (enc3): Conv2d(16, 32, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1))  
  (enc4): Conv2d(32, 64, kernel_size=(4, 4), stride=(2, 2))  
  (fc1): Linear(in_features=64, out_features=128, bias=True)  
  (fc_mu): Linear(in_features=128, out_features=2, bias=True) Latent space dimension = 2  
  (fc_log_var): Linear(in_features=128, out_features=2, bias=True)  
  (fc2): Linear(in_features=2, out_features=64, bias=True)  
  (dec1): ConvTranspose2d(64, 64, kernel_size=(4, 4), stride=(1, 1))  
  (dec2): ConvTranspose2d(64, 32, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1))  
  (dec3): ConvTranspose2d(32, 16, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1))  
  (dec4): ConvTranspose2d(16, 3, kernel_size=(4, 4), stride=(2, 2), padding=(1, 1))  
  (classifier): Linear(in_features=2, out_features=6, bias=True)  
)
```

## Latent space



## Sampling

