## **## Module 12 Assignment: Autoencoders**

Video 13.1

https://www.youtube.com/watch?v=kIGHE7Cfe1s

Video 13.2

https://www.youtube.com/watch?v=Rm9bJcDd1KU

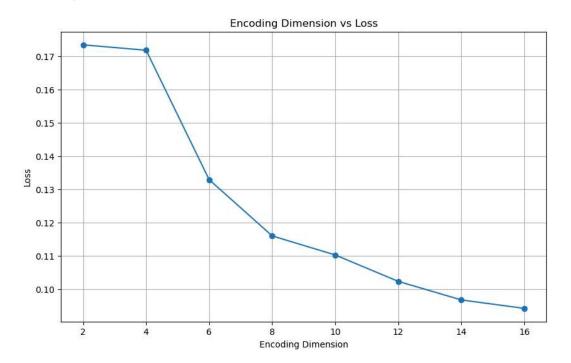
Video 13.3

https://youtu.be/6HjZk-3LsjE

1. Change the `encoding\_dim` through various values (`range(2,18,2)` and save the loss you can get. Plot the 8 pairs of dimensions vs loss on a scatter plot

```
from keras.layers import Input, Dense
In [16]:
             from keras.models import Model
             from keras.datasets import mnist
             import numpy as np
             import matplotlib.pyplot as plt
             # Loading MNIST dataset
             (xtrain, ytrain), (xtest, ytest) = mnist.load_data()
             xtrain = xtrain.astype('float32') / 255.
             xtest = xtest.astype('float32') / 255.
             xtrain = xtrain.reshape((len(xtrain), np.prod(xtrain.shape[1:])))
             xtest = xtest.reshape((len(xtest), np.prod(xtest.shape[1:])))
             xtrain.shape, xtest.shape
             # Creating function to train the autoencoder and return the validation
             def train_autoencoder(encoding_dim):
                 input img = Input(shape=(784,))
                 x = Dense(256, activation='relu')(input_img)
                 x = Dense(128, activation='relu')(x)
                 encoded = Dense(encoding dim, activation='relu')(x)
                 x = Dense(128, activation='relu')(encoded)
                 x = Dense(256, activation='relu')(x)
                 decoded = Dense(784, activation='sigmoid')(x)
                 autoencoder = Model(input_img, decoded)
                 autoencoder.compile(optimizer='adam', loss='binary_crossentropy')
                 history = autoencoder.fit(xtrain, xtrain,
                                           epochs=50,
                                           batch size=256,
                                           shuffle=True,
                                           validation_data=(xtest, xtest),
                                           verbose=0)
                 val_loss = history.history['val_loss'][-1]
                 return val_loss
             # Training autoencoders with various encoding dimensions and saving th
             dimensions = range(2, 18, 2)
             losses = []
             for encoding_dim in dimensions:
                 loss = train_autoencoder(encoding_dim)
                 losses.append(loss)
                 print(f'Encoding Dim: {encoding_dim}, Loss: {loss}')
             # Plotting the losses vs encoding dimensions
             plt.figure(figsize=(10, 6))
             plt.plot(dimensions, losses, marker='o')
             plt.title('Encoding Dimension vs Loss')
             plt.xlabel('Encoding Dimension')
             plt.ylabel('Loss')
             plt.grid(True)
             plt.show()
```

```
Encoding Dim: 2, Loss: 0.1735038012266159
Encoding Dim: 4, Loss: 0.1718769073486328
Encoding Dim: 6, Loss: 0.13293473422527313
Encoding Dim: 8, Loss: 0.11601827293634415
Encoding Dim: 10, Loss: 0.11028366535902023
Encoding Dim: 12, Loss: 0.10233931243419647
Encoding Dim: 14, Loss: 0.09675975143909454
Encoding Dim: 16, Loss: 0.09421230852603912
```



2. **After** training an autoencoder with encoding\_dim=8, apply noise (like the previous assignment) to *only* the input of the trained autoencoder (not the output). The output images should be without noise.

Print a few noisy images along with the output images to show they don't have noise.

```
In [15]: ▶ # Training an autoencoder with encoding dim=8
             encoding dim = 8
             input_img = Input(shape=(784,))
             x = Dense(256, activation='relu')(input_img)
             x = Dense(128, activation='relu')(x)
             encoded = Dense(encoding dim, activation='relu')(x)
             x = Dense(128, activation='relu')(encoded)
             x = Dense(256, activation='relu')(x)
             decoded = Dense(784, activation='sigmoid')(x)
             autoencoder = Model(input_img, decoded)
             encoder = Model(input img, encoded)
             encoded input = Input(shape=(encoding dim,))
             dcd1 = autoencoder.layers[-1]
             dcd2 = autoencoder.layers[-2]
             dcd3 = autoencoder.layers[-3]
             decoder = Model(encoded_input, dcd1(dcd2(dcd3(encoded_input))))
             autoencoder.compile(optimizer='adam', loss='binary crossentropy')
             autoencoder.fit(xtrain, xtrain,
                             epochs=50,
                             batch_size=256,
                             shuffle=True,
                             validation data=(xtest, xtest),
                             verbose=0)
             # Adding noise to the input
             noise_factor = 0.5
             xtest_noisy = xtest + noise_factor * np.random.normal(loc=0.0, scale=1
             xtest noisy = np.clip(xtest noisy, 0., 1.)
             # Getting the denoised images
             decoded_imgs = autoencoder.predict(xtest_noisy)
             # Displaying the noisy and denoised images
             n = 10 # number of digits to display
             plt.figure(figsize=(20, 4))
             for i in range(n):
                 # Display original + noise
                 ax = plt.subplot(2, n, i + 1)
                 plt.imshow(xtest_noisy[i].reshape(28, 28))
                 plt.gray()
                 ax.get_xaxis().set_visible(False)
                 ax.get yaxis().set visible(False)
                 # Display denoised image
                 ax = plt.subplot(2, n, i + 1 + n)
                 plt.imshow(decoded_imgs[i].reshape(28, 28))
                 plt.grav()
                 ax.get xaxis().set visible(False)
                 ax.get_yaxis().set_visible(False)
```

plt.show() 313/313 -**- 2s** 5ms/step 7210119475 3 2 3 3 3 3 3 3 3