from keras.layers import Input, Dense

from keras.models import Model

from keras.datasets import fashion\_mnist

import numpy as np

# this is the size of our encoded representations

encoding\_dim = 32  # 32 floats -> compression factor 24.5, assuming the input is 784 floats

# this is our input placeholder

input\_img = Input(shape=(784,))

# "encoded" is the encoded representation of the input

encoded = Dense(encoding\_dim, activation='relu')(input\_img)

# "decoded" is the lossy reconstruction of the input

decoded = Dense(784, activation='sigmoid')(encoded)

# this model maps an input to its reconstruction

autoencoder = Model(input\_img, decoded)

# this model maps an input to its encoded representation

autoencoder.compile(optimizer='adadelta', loss='binary\_crossentropy', metrics=['accuracy'])

(x\_train, y\_train), (x\_test, y\_test) = fashion\_mnist.load\_data()

x\_train = x\_train.astype('float32') / 255.

x\_test = x\_test.astype('float32') / 255.

x\_train = x\_train.reshape((len(x\_train), np.prod(x\_train.shape[1:])))

x\_test = x\_test.reshape((len(x\_test), np.prod(x\_test.shape[1:])))

autoencoder.fit(x\_train, x\_train,

                epochs=5,

                batch\_size=256,

                shuffle=True,

                validation\_data=(x\_test, x\_test))

A screenshot of a computer

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from keras.layers import Input, Dense

from keras.models import Model

# This is the size of our encoded representation

encoding\_dim = 32  # 32 floats -> compression of factor 24.5, assuming the input is 784 floats

# This is our input placeholder

input\_img = Input(shape=(784,))

# "encoded" is the encoded representation of the input

encoded1 = Dense(128, activation='relu')(input\_img)

encoded2 = Dense(encoding\_dim, activation='relu')(encoded1)

# "decoded" is the lossy reconstruction of the input

decoded1 = Dense(128, activation='relu')(encoded2)

decoded2 = Dense(784, activation='sigmoid')(decoded1)

# This model maps an input to its reconstruction

autoencoder = Model(input\_img, decoded2)

# This model maps an input to its encoded representation

encoder = Model(input\_img, encoded2)

# This is our decoder model

encoded\_input = Input(shape=(encoding\_dim,))

decoder\_layer1 = autoencoder.layers[-2]

decoder\_layer2 = autoencoder.layers[-1]

decoder = Model(encoded\_input, decoder\_layer2(decoder\_layer1(encoded\_input)))

# Compile the model

autoencoder.compile(optimizer='adadelta', loss='binary\_crossentropy',metrics ='accuracy')

# Load the MNIST dataset

from keras.datasets import mnist, fashion\_mnist

import numpy as np

(x\_train, y\_train), (x\_test, y\_test) = fashion\_mnist.load\_data()

# Normalize and flatten the data

x\_train = x\_train.astype('float32') / 255.

x\_test = x\_test.astype('float32') / 255.

x\_train = x\_train.reshape((len(x\_train), np.prod(x\_train.shape[1:])))

x\_test = x\_test.reshape((len(x\_test), np.prod(x\_test.shape[1:])))

# Train the autoencoder

autoencoder.fit(x\_train, x\_train,

                epochs=5,

                batch\_size=256,

                shuffle=True,

                validation\_data=(x\_test, x\_test))

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import matplotlib.pyplot as plt

# Get the reconstructed images for the test set

reconstructed\_imgs = autoencoder.predict(x\_test)

# Choose a random image from the test set

n = 10  # index of the image to be plotted

plt.figure(figsize=(10, 5))

# Plot the original image

ax = plt.subplot(1, 2, 1)

plt.imshow(x\_test[n].reshape(28, 28))

plt.gray()

ax.get\_xaxis().set\_visible(False)

ax.get\_yaxis().set\_visible(False)

ax.set\_title("Original Image")

# Plot the reconstructed image

ax = plt.subplot(1, 2, 2)

plt.imshow(reconstructed\_imgs[n].reshape(28, 28))

plt.gray()

ax.get\_xaxis().set\_visible(False)

ax.get\_yaxis().set\_visible(False)

ax.set\_title("Reconstructed Image")

plt.show()

A collage of images of a person

Description automatically generated

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from keras.models import Model

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encoded = Dense(encoding\_dim, activation='relu')(input\_img)

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autoencoder.compile(optimizer='adadelta', loss='binary\_crossentropy',metrics ='accuracy')

from keras.datasets import fashion\_mnist

import numpy as np

(x\_train, \_), (x\_test, \_) = fashion\_mnist.load\_data()

x\_train = x\_train.astype('float32') / 255.

x\_test = x\_test.astype('float32') / 255.

x\_train = x\_train.reshape((len(x\_train), np.prod(x\_train.shape[1:])))

x\_test = x\_test.reshape((len(x\_test), np.prod(x\_test.shape[1:])))

#introducing noise

noise\_factor = 0.5

x\_train\_noisy = x\_train + noise\_factor \* np.random.normal(loc=0.0, scale=1.0, size=x\_train.shape)

x\_test\_noisy = x\_test + noise\_factor \* np.random.normal(loc=0.0, scale=1.0, size=x\_test.shape)

autoencoder.fit(x\_train\_noisy, x\_train,

                epochs=10,

                batch\_size=256,

                shuffle=True,

                validation\_data=(x\_test\_noisy, x\_test\_noisy))

A screenshot of a computer

Description automatically generated

import matplotlib.pyplot as plt

# Get the reconstructed images for the test set

reconstructed\_imgs = autoencoder.predict(x\_test\_noisy)

# Choose a random image from the test set

n = 10  # index of the image to be plotted

plt.figure(figsize=(10, 5))

# Plot the original noisy image

ax = plt.subplot(1, 2, 1)

plt.imshow(x\_test\_noisy[n].reshape(28, 28))

plt.gray()

ax.get\_xaxis().set\_visible(False)

ax.get\_yaxis().set\_visible(False)

ax.set\_title("Noisy Image")

# Plot the reconstructed image

ax = plt.subplot(1, 2, 2)

plt.imshow(reconstructed\_imgs[n].reshape(28, 28))

plt.gray()

ax.get\_xaxis().set\_visible(False)

ax.get\_yaxis().set\_visible(False)

ax.set\_title("Reconstructed Image")

plt.show()

A screenshot of a screen

Description automatically generated

import matplotlib.pyplot as plt

# Train the autoencoder

history = autoencoder.fit(x\_train\_noisy, x\_train,

                epochs=10,

                batch\_size=256,

                shuffle=True,

                validation\_data=(x\_test\_noisy, x\_test\_noisy))

# Plot the loss

plt.plot(history.history['loss'], label='train')

plt.plot(history.history['val\_loss'], label='test')

plt.title('Model Loss')

plt.ylabel('Loss')

plt.xlabel('Epoch')

plt.legend()

plt.show()

# Plot the accuracy

plt.plot(history.history['accuracy'], label='train')

plt.plot(history.history['val\_accuracy'], label='test')

plt.title('Model Accuracy')

plt.ylabel('Accuracy')

plt.xlabel('Epoch')

plt.legend()

plt.show()

A graph with a line and numbers

Description automatically generated with medium confidence

A graph with a line and a line

Description automatically generated with medium confidence