



Dinesh Madhav RollNo-29 DeepLearning BE-CSE(DS)

```
import keras
from keras import layers
from keras.datasets import mnist
import numpy as np

(x_train, _), (x_test, _) = mnist.load_data()

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
11490434/11490434 [=====] - 0s 0us/step

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:])))
print(x_train.shape)
print(x_test.shape)

(60000, 784)
(10000, 784)

encoding_dim = 32
input_img= keras.Input(shape=(784,))
encoded = layers.Dense(encoding_dim, activation='relu')(input_img)
decoded = layers.Dense(784, activation='sigmoid')(encoded)
autoencoder=keras.Model(input_img,decoded)

encoder = keras.Model(input_img, encoded)

encoded_input= keras.Input(shape=(encoding_dim,))
decoder_layer=autoencoder.layers[-1]
decoder = keras.Model(encoded_input, decoder_layer(encoded_input))

autoencoder.compile(optimizer='adam',loss='binary_crossentropy')

autoencoder.fit(x_train, x_train,
                epochs=20,
                batch_size=256,
                shuffle=True,
                validation_data=(x_test, x_test))

Epoch 1/20
235/235 [=====] - 2s 8ms/step - loss: 0.0926 - val_loss: 0.0915
Epoch 2/20
235/235 [=====] - 1s 4ms/step - loss: 0.0926 - val_loss: 0.0914
Epoch 3/20
235/235 [=====] - 1s 4ms/step - loss: 0.0925 - val_loss: 0.0915
Epoch 4/20
235/235 [=====] - 1s 4ms/step - loss: 0.0925 - val_loss: 0.0914
Epoch 5/20
235/235 [=====] - 1s 4ms/step - loss: 0.0925 - val_loss: 0.0914
Epoch 6/20
235/235 [=====] - 1s 4ms/step - loss: 0.0925 - val_loss: 0.0914
Epoch 7/20
235/235 [=====] - 1s 4ms/step - loss: 0.0925 - val_loss: 0.0914
Epoch 8/20
235/235 [=====] - 1s 4ms/step - loss: 0.0925 - val_loss: 0.0914
Epoch 9/20
235/235 [=====] - 1s 4ms/step - loss: 0.0925 - val_loss: 0.0913
Epoch 10/20
235/235 [=====] - 1s 6ms/step - loss: 0.0925 - val_loss: 0.0914
Epoch 11/20
235/235 [=====] - 1s 6ms/step - loss: 0.0925 - val_loss: 0.0915
Epoch 12/20
235/235 [=====] - 1s 4ms/step - loss: 0.0925 - val_loss: 0.0913
Epoch 13/20
235/235 [=====] - 1s 4ms/step - loss: 0.0924 - val_loss: 0.0913
Epoch 14/20
235/235 [=====] - 1s 4ms/step - loss: 0.0925 - val_loss: 0.0914
Epoch 15/20
235/235 [=====] - 1s 4ms/step - loss: 0.0924 - val_loss: 0.0914
Epoch 16/20
235/235 [=====] - 1s 4ms/step - loss: 0.0924 - val_loss: 0.0913
Epoch 17/20
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235/235 [=====] - 1s 4ms/step - loss: 0.0924 - val_loss: 0.0913
Epoch 18/20
235/235 [=====] - 1s 4ms/step - loss: 0.0924 - val_loss: 0.0914
Epoch 19/20
235/235 [=====] - 1s 5ms/step - loss: 0.0924 - val_loss: 0.0912
Epoch 20/20
235/235 [=====] - 1s 4ms/step - loss: 0.0924 - val_loss: 0.0913
<keras.callbacks.History at 0x7a4fc028ff70>

# Encode and decode some digits
# Note that we take them from the *test* set
encoded_imgs = encoder.predict(x_test)
decoded_imgs = decoder.predict(encoded_imgs)

313/313 [=====] - 1s 2ms/step
313/313 [=====] - 0s 1ms/step

# Use Matplotlib (don't ask)
import matplotlib.pyplot as plt

n = 10 # How many digits we will display
plt.figure(figsize=(20, 4))
for i in range(n):
    # Display original
    ax = plt.subplot(2, n, i + 1)
    plt.imshow(x_test[i].reshape(28, 28))
    plt.gray()
    ax.get_xaxis().set_visible(False)
    ax.get_yaxis().set_visible(False)

    # Display reconstruction
    ax = plt.subplot(2, n, i + 1 + n)
    plt.imshow(decoded_imgs[i].reshape(28, 28))
    plt.gray()
    ax.get_xaxis().set_visible(False)
    ax.get_yaxis().set_visible(False)
plt.show()

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