

NEURAL NETWORK DEEP LEARNING

ICP 6

700758238

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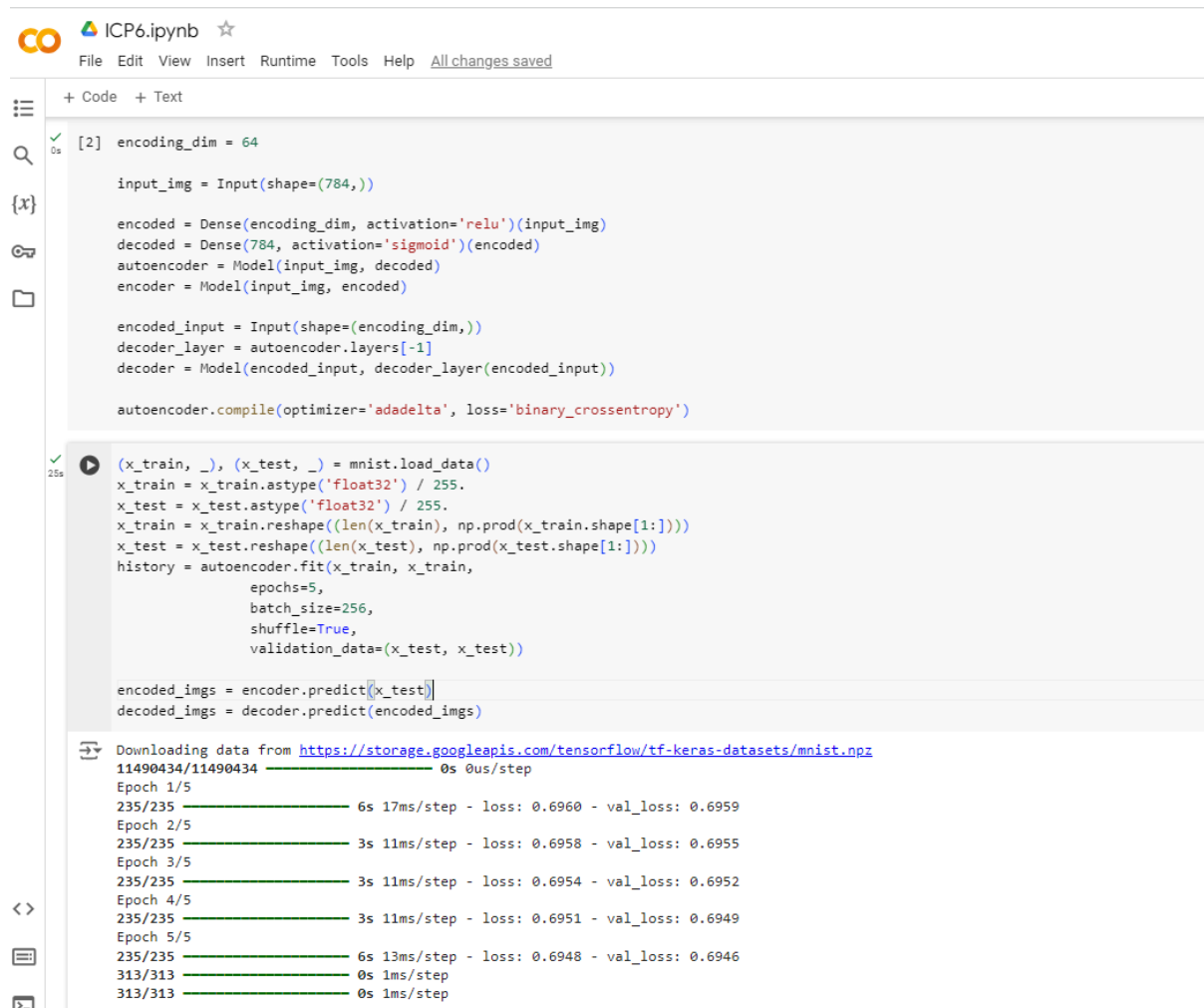
GitHub:

Repository URL for the source code:

https://github.com/axk82380/NNPL/tree/main/ICP_6

Video Link:

<https://drive.google.com/file/d/1dOJwO20nMdCU-MkfSiyul6BAOE8rIpXD/view?usp=sharing>



```
ICP6.ipynb
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[2] encoding_dim = 64

input_img = Input(shape=(784,))

encoded = Dense(encoding_dim, activation='relu')(input_img)
decoded = Dense(784, activation='sigmoid')(encoded)
autoencoder = Model(input_img, decoded)
encoder = Model(input_img, encoded)

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

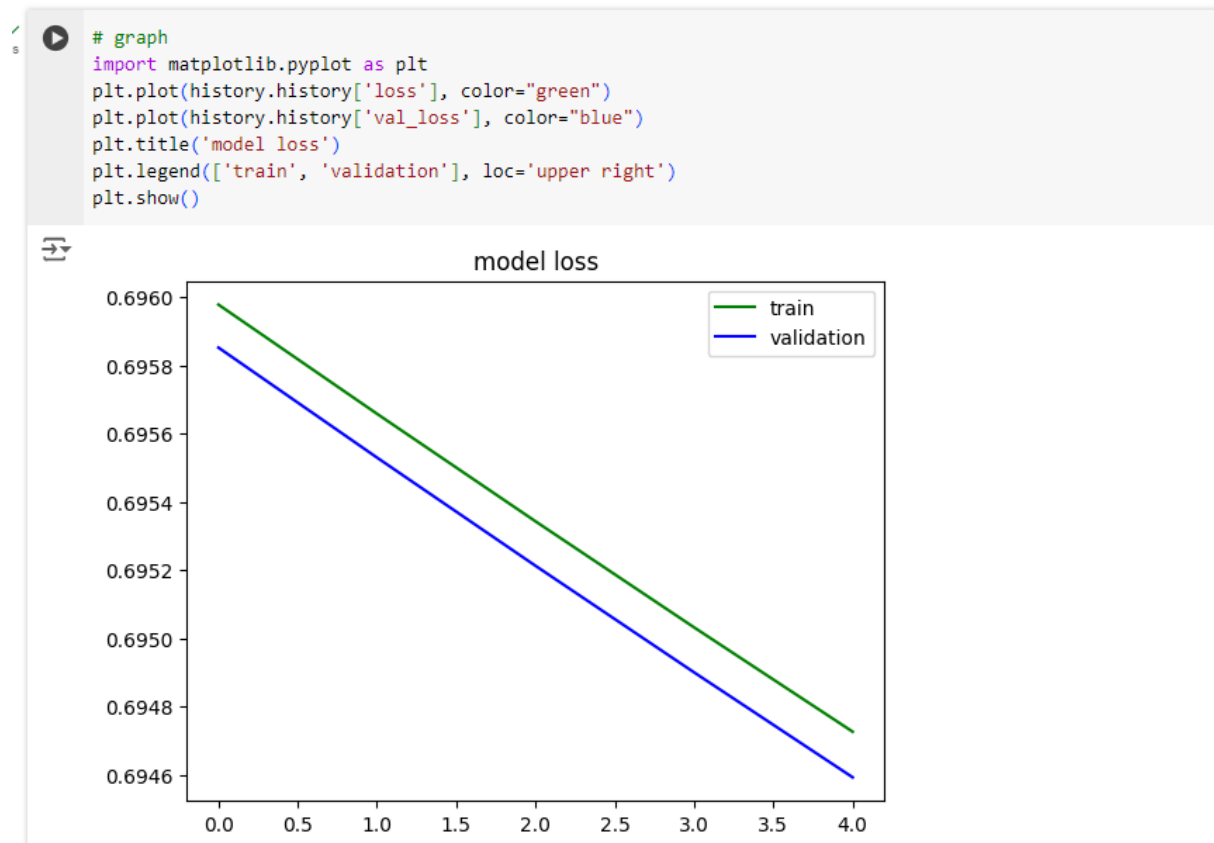
autoencoder.compile(optimizer='adadelta', loss='binary_crossentropy')

(x_train, _), (x_test, _) = 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:])))
history = autoencoder.fit(x_train, x_train,
                        epochs=5,
                        batch_size=256,
                        shuffle=True,
                        validation_data=(x_test, x_test))

encoded_imgs = encoder.predict(x_test)
decoded_imgs = decoder.predict(encoded_imgs)

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
11490434/11490434 — 0s 0us/step
Epoch 1/5
235/235 — 6s 17ms/step - loss: 0.6960 - val_loss: 0.6959
Epoch 2/5
235/235 — 3s 11ms/step - loss: 0.6958 - val_loss: 0.6955
Epoch 3/5
235/235 — 3s 11ms/step - loss: 0.6954 - val_loss: 0.6952
Epoch 4/5
235/235 — 3s 11ms/step - loss: 0.6951 - val_loss: 0.6949
Epoch 5/5
235/235 — 6s 13ms/step - loss: 0.6948 - val_loss: 0.6946
313/313 — 0s 1ms/step
313/313 — 0s 1ms/step
```

Autoencoder without hidden layer.



Graph for validation and training.



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```
[5] input_size = 784
    hidden_size = 128
    code_size = 32

    input_img = Input(shape=(input_size,))
    hidden_1 = Dense(hidden_size, activation='relu')(input_img)
    code = Dense(code_size, activation='relu')(hidden_1)
    hidden_2 = Dense(hidden_size, activation='relu')(code)
    output_img = Dense(input_size, activation='sigmoid')(hidden_2)

    autoencoder = Model(input_img, output_img)
    autoencoder.compile(optimizer='adam', loss='binary_crossentropy')
```

```
[6] (x_train, _), (x_test, _) = 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:])))
    history = autoencoder.fit(x_train, x_train,
                             epochs=5,
                             batch_size=256,
                             shuffle=True,
                             validation_data=(x_test, x_test))
```



```
Epoch 1/5
235/235 — 6s 18ms/step - loss: 0.3247 - val_loss: 0.1528
Epoch 2/5
235/235 — 4s 14ms/step - loss: 0.1427 - val_loss: 0.1200
Epoch 3/5
235/235 — 6s 19ms/step - loss: 0.1187 - val_loss: 0.1097
Epoch 4/5
235/235 — 4s 14ms/step - loss: 0.1090 - val_loss: 0.1035
Epoch 5/5
235/235 — 5s 15ms/step - loss: 0.1039 - val_loss: 0.0998
```

Autoencoder with hidden layer. Here the validation loss is 0.0998.



ICP6.ipynb ☆

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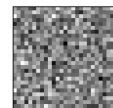
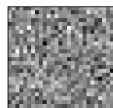
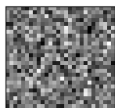
```
[7] encoded_imgs = encoder.predict(x_test)
    decoded_imgs = decoder.predict(encoded_imgs)

    import matplotlib.pyplot as plt

    n = 3
    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()
```

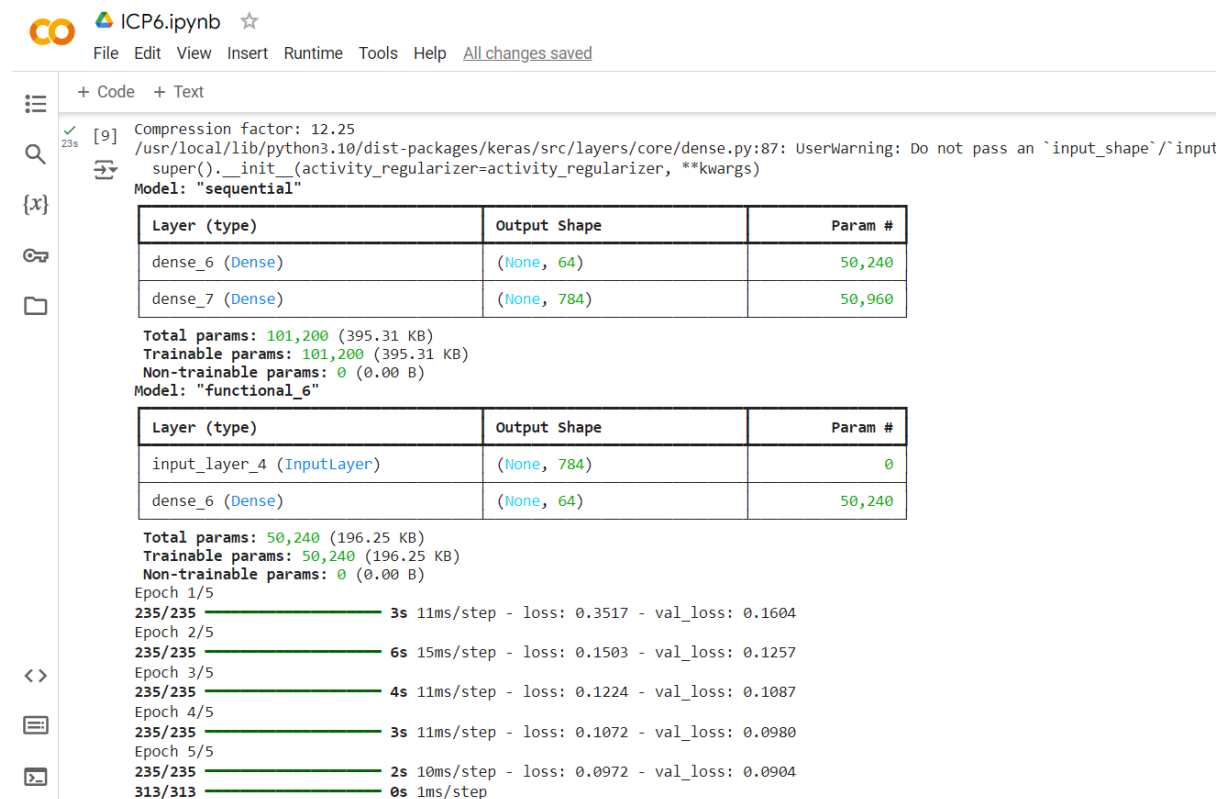
```
313/313 — 0s 1ms/step
313/313 — 0s 1ms/step
```



Printing original and reconstructed images



Printing model loss after adding hidden layer.



Computing and printing compression factor, and validation loss.