



# Neural Networks ASSIGNMENT 4

Yamini Radha Veeranki

700741751

```
# Predict on test data
decoded_imgs = autoencoder.predict(x_test)

# Visualize reconstructed image and original image
import matplotlib.pyplot as plt

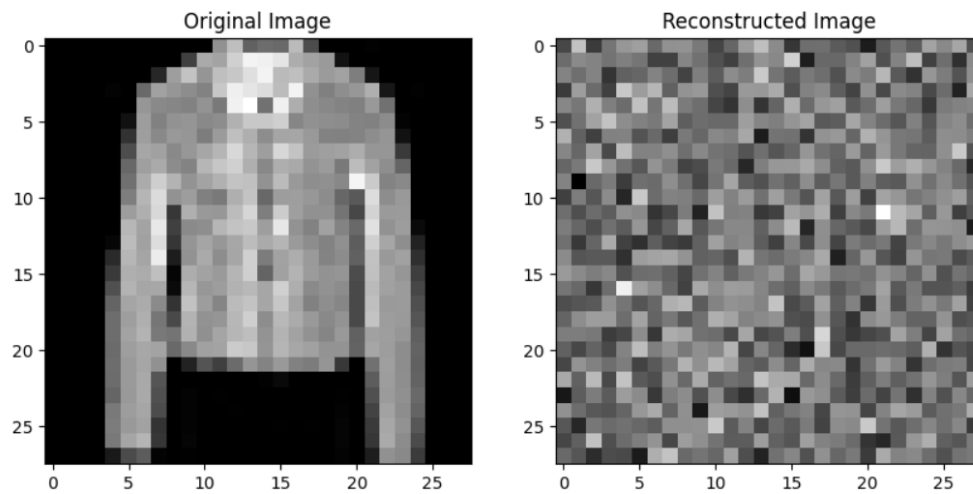
# Choose an index of a test image to visualize
idx = 10

# Reshape the test image
test_img = x_test[idx].reshape(28, 28)

# Reshape the reconstructed image
reconstructed_img = decoded_imgs[idx].reshape(28, 28)

# Plot the original and reconstructed images side by side
plt.figure(figsize=(10, 5))
plt.subplot(1, 2, 1)
plt.imshow(test_img, cmap='gray')
plt.title('Original Image')
plt.subplot(1, 2, 2)
plt.imshow(reconstructed_img, cmap='gray')
plt.title('Reconstructed Image')
plt.show()
```

Epoch 1/20



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```
In [10]: from keras.layers import Input, Dense
         from keras.models import Model

         # this is the size of our encoded representations
         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
         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'])

         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)

         history=autoencoder.fit(x_train_noisy, x_train,
                                epochs=10,
                                batch_size=256,
                                shuffle=True,
                                validation_data=(x_test_noisy, x_test_noisy))
```

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```
# Get the reconstructed images
reconstructed_imgs = autoencoder.predict(x_test_noisy)

# Select one image to display
img_to_display = 0

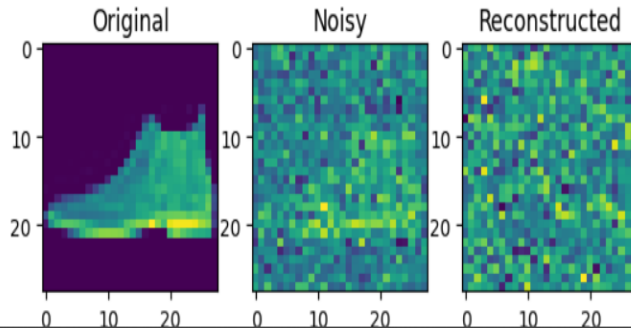
# Display the original, noisy, and reconstructed images side by side
plt.subplot(1, 3, 1)
plt.imshow(x_test[img_to_display].reshape(28, 28))
plt.title('Original')

plt.subplot(1, 3, 2)
plt.imshow(x_test_noisy[img_to_display].reshape(28, 28))
plt.title('Noisy')

plt.subplot(1, 3, 3)
plt.imshow(reconstructed_imgs[img_to_display].reshape(28, 28))
plt.title('Reconstructed')

plt.show()
```

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



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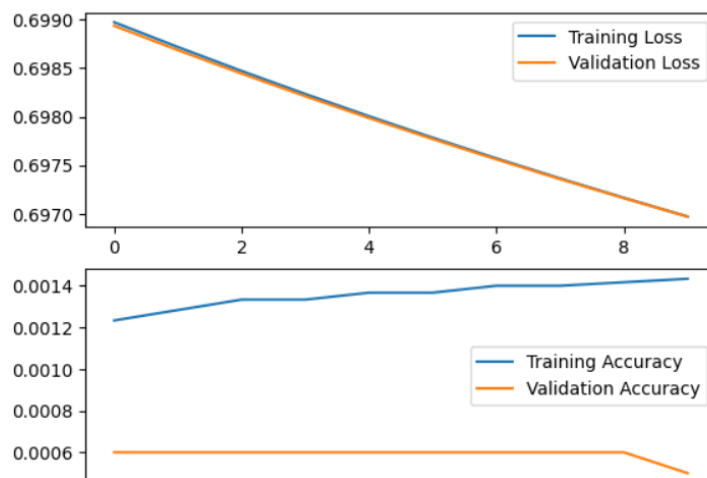
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```
plt.subplot(2, 1, 1)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.legend()

plt.subplot(2, 1, 2)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.legend()

plt.show()
```



**GITHUB LINK:** [https://github.com/Yaminiradha/NN\\_700741751\\_Assignment4](https://github.com/Yaminiradha/NN_700741751_Assignment4)

**VIDEO\_LINK:**

[https://drive.google.com/file/d/1JE9odnltuXdQKgDI0a8V\\_LkPxbTqoPVf/view?usp=drive\\_link](https://drive.google.com/file/d/1JE9odnltuXdQKgDI0a8V_LkPxbTqoPVf/view?usp=drive_link)