## Spring 2025: Neural Networks & Deep Learning - ICP -6

## Assignment -6

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Github Link: https://github.com/VanithaChintalapudi10/Neural-network-deep-learning

## Video Link:

https://drive.google.com/file/d/10mOJQLiMaBG7ujRWbXzSZLCl6webni1/view?usp=drive\_link

- 1. Add one more hidden layer to autoencoder
- 2. Do the prediction on the test data and then visualize one of the reconstructed version of that test data. Also, visualize the same test data before reconstruction using Matplotlib
- 3. Repeat the question 2 on the denoisening autoencoder
- 4. plot loss and accuracy using the history object

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                                                                                                                                                                                                                                              from keras.layers import Input, Dense from keras.aodels import Model from keras.addasets import fashion_mnist
                 import numpy as np
import matplotlib.pyplot as plt
<>
              # === Load data ===
(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), pp.prod(x_train.shape[1:])))
x_test = x_test.reshape((len(x_test), np.prod(x_test.shape[1:])))
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# === 1. Autoencoder with extra hidden layers ===
                    # === 1. Autoencoder with extra floden layers ===
input(shape=(784,))
encoded = Dense(128, activation='relu')(input_img)
encoded = Dense(32, activation='relu')(encoded) # bottleneck
                    decoded = Dense(128, activation='relu')(encoded)
decoded = Dense(784, activation='sigmoid')(decoded)
                   autoencoder = Model(input_img, decoded)
autoencoder.compile(optimizer='adam', loss='binary_crossentropy')
                   history = autoencoder.fit(x_train, x_train,
              epochs=15,
batch_size=256,
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                                    batch size=256,
                                                                                                                                                    i≡ m O
                                     validation_data=(x_test, x_test))
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            # === 2. Predict & visualize ===
            decoded_imgs = autoencoder.predict(x_test)
<>
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            plt.figure(figsize=(10, 4))
             for i in range(n):
# Original
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               " on gainer
ax = plt.subplot(2, n, i + 1)
plt.imshow(x_test[i].reshape(28, 28), cmap='gray')
plt.title("Original")
plt.axis('off')
# Reconstructed
                " **econstructed
ax = plt.subplot(2, n, i + 1 + n)
plt.inshow(decoded_imgs[i].reshape(28, 28), cmap='gray')
plt.title("*Reconstructed")
plt.axis('off')
            plt.tight_layout()
plt.show()
            # === 3. Denoising Autoencoder ===
            # Add noise
            # Add 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)

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                      ax = plt.subplot(3, n, i + 1)
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                     plt.imshow(x_test_noisy[i].reshape(28, 28), cmap='gray')
                      plt.title("Noisy")
                     plt.axis('off')
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                      # Denoised
<>
                      ax = plt.subplot(3, n, i + 1 + n)
                      plt.imshow(denoised_imgs[i].reshape(28, 28), cmap='gray')
{x}
                      plt.title("Denoised")
                     plt.axis('off')
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                      # Ground truth
                      ax = plt.subplot(3, n, i + 1 + 2 * n)
                      plt.imshow(x_test[i].reshape(28, 28), cmap='gray')
                      plt.title("Original")
                      plt.axis('off')
                plt.tight_layout()
                plt.show()
                 # === 4. Plot training/validation loss ===
                 plt.plot(history.history['loss'], label='Train Loss - AE')
                plt.plot(history.history['val_loss'], label='Val_Loss - AE')
plt.plot(denoise_history.history['loss'], label='Train Loss - DAE')
                 plt.plot(denoise_history.history['val_loss'], label='Val Loss - DAE')
                 plt.title("Loss over Epochs")
                plt.xlabel("Epoch")
plt.ylabel("Loss")
                 plt.legend()
                 plt.grid(True)
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```

plt.show()



