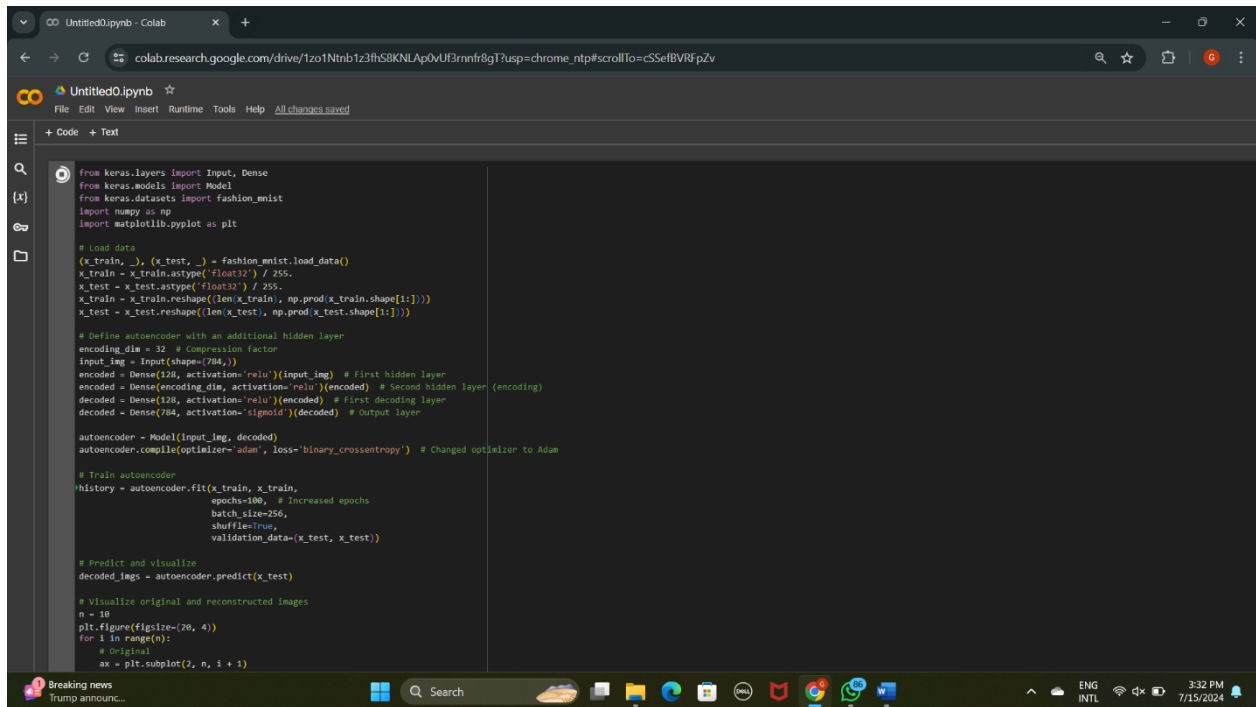


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Github Link: [GaneshKumarKorra/ICP-5 \(github.com\)](https://github.com/GaneshKumarKorra/ICP-5)



```
from keras.layers import Input, Dense
from keras.models import Model
from keras.datasets 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), np.prod(x_train.shape[1:])))
x_test = x_test.reshape((len(x_test), np.prod(x_test.shape[1:])))

# Define autoencoder with an additional hidden layer
encoding_dim = 32 # Compression factor
input_img = Input(shape=(784,))
encoded = Dense(128, activation='relu')(input_img) # First hidden layer
encoded = Dense(encoding_dim, activation='relu')(encoded) # Second hidden layer (encoding)
decoded = Dense(128, activation='relu')(encoded) # First decoding layer
decoded = Dense(784, activation='sigmoid')(decoded) # Output layer

autoencoder = Model(input_img, decoded)
autoencoder.compile(optimizer='adam', loss='binary_crossentropy') # Changed optimizer to Adam

# Train autoencoder
history = autoencoder.fit(x_train, x_train,
                          epochs=100, # Increased epochs
                          batch_size=256,
                          shuffle=True,
                          validation_data=(x_test, x_test))

# Predict and visualize
decoded_imgs = autoencoder.predict(x_test)

# Visualize original and reconstructed images
n = 10
plt.figure(figsize=(20, 4))
for i in range(n):
    # Original
    ax = plt.subplot(2, n, i + 1)
```

```
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# Predict and visualize denoised images
denoised_imgs = denoising_autoencoder.predict(x_test_noisy)

plt.figure(figsize=(20, 4))
for i in range(n):
    # Noisy image
    ax = plt.subplot(3, n, i + 1)
    plt.imshow(x_test_noisy[i].reshape(28, 28))
    plt.gray()
    ax.get_xaxis().set_visible(False)
    ax.get_yaxis().set_visible(False)

    # Reconstructed image
    ax = plt.subplot(3, n, i + 1 + n)
    plt.imshow(denoised_imgs[i].reshape(28, 28))
    plt.gray()
    ax.get_xaxis().set_visible(False)
    ax.get_yaxis().set_visible(False)
plt.show()

# Plot loss and accuracy
plt.figure(figsize=(12, 4))
plt.subplot(1, 2, 1)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Training and Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()

plt.subplot(1, 2, 2)
plt.plot(history.denoising_history['loss'], label='Denoising Training Loss')
plt.plot(history.denoising_history['val_loss'], label='Denoising Validation Loss')
plt.title('Denoising Training and Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()

plt.show()

... 235/235 [=====] - 1s 5ms/step - loss: 0.2768 - val_loss: 0.2800
Epoch: 73/100

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```

```
CO Untitled0.ipynb - Colab
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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)

# Reconstructed
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()

# Implement a denoising autoencoder
noise_factor = 0.2 # Reduced noise factor
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)
x_train_noisy = np.clip(x_train_noisy, 0., 1.)
x_test_noisy = np.clip(x_test_noisy, 0., 1.)

# Define autoencoder for denoising
input_img_noisy = Input(shape=(784,))
encoded_noisy = Dense(128, activation='relu')(input_img_noisy)
encoded_noisy = Dense(encoding_dim, activation='relu')(encoded_noisy)
decoded_noisy = Dense(128, activation='relu')(encoded_noisy)
decoded_noisy = Dense(784, activation='sigmoid')(decoded_noisy)

denoising_autoencoder = Model(input_img_noisy, decoded_noisy)
denoising_autoencoder.compile(optimizer='adam', loss='binary_crossentropy') # Changed optimizer to Adam

# Train denoising autoencoder
history_denoising = denoising_autoencoder.fit(x_train_noisy, x_train,
                                             epochs=100, # Increased epochs
                                             batch_size=256,
                                             shuffle=True,
                                             validation_data=(x_test_noisy, x_test))

# Predict and visualize denoised images
denoised_imgs = denoising_autoencoder.predict(x_test_noisy)

plt.figure(figsize=(20, 4))
```

```
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# Predict and visualize denoised images
denoised_imgs = denoising_autoencoder.predict(x_test_noisy)

plt.figure(figsize=(20, 4))
for i in range(5):
    # Noisy image
    ax = plt.subplot(3, 5, i + 1)
    plt.imshow(x_test_noisy[i].reshape(28, 28))
    plt.gray()
    ax.get_xaxis().set_visible(False)
    ax.get_yaxis().set_visible(False)

    # Reconstructed image
    ax = plt.subplot(3, 5, i + 1 + 5)
    plt.imshow(denoised_imgs[i].reshape(28, 28))
    plt.gray()
    ax.get_xaxis().set_visible(False)
    ax.get_yaxis().set_visible(False)
    plt.show()

# Plot loss and accuracy
plt.figure(figsize=(12, 4))
plt.subplot(1, 2, 1)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Training and Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()

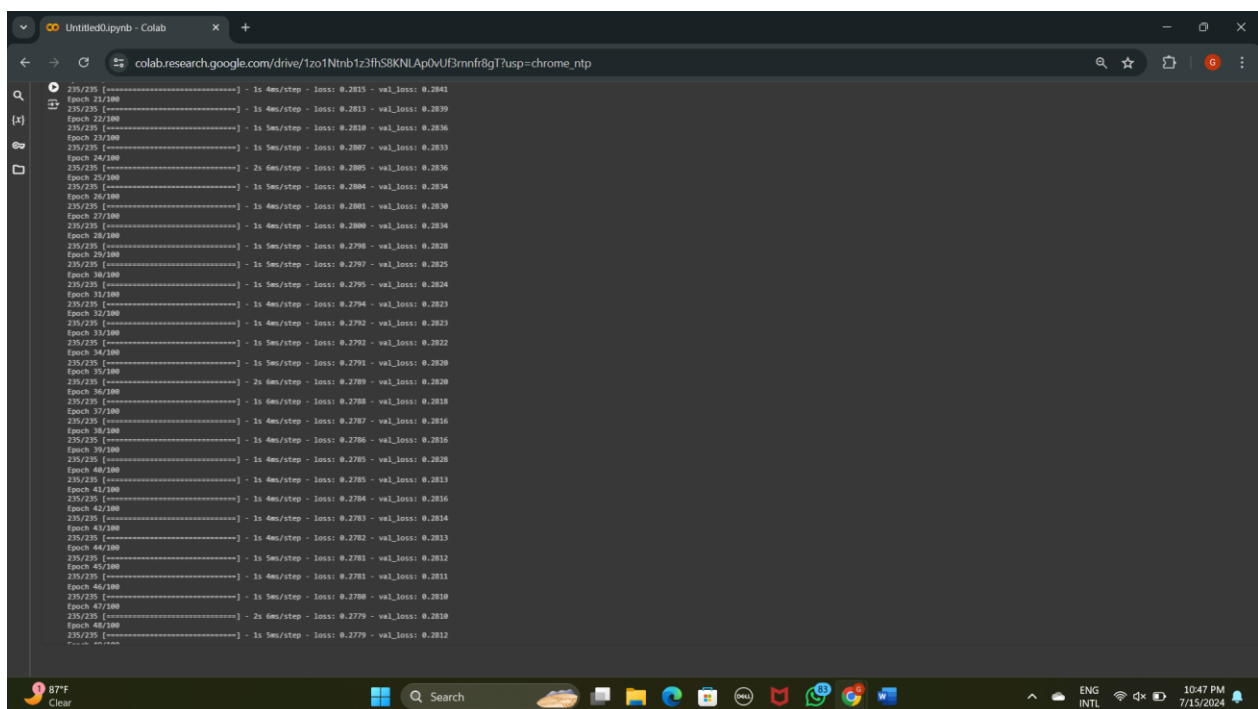
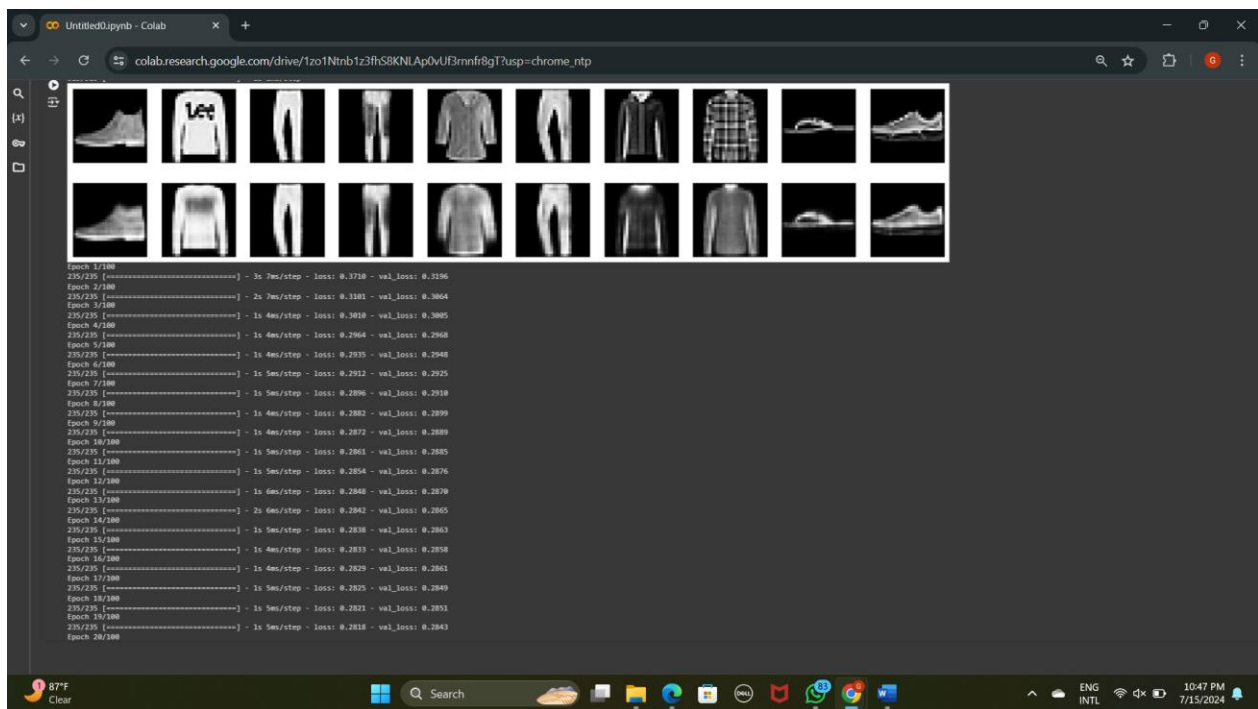
plt.subplot(1, 2, 2)
plt.plot(history.denoising_history['loss'], label='Denoising Training Loss')
plt.plot(history.denoising_history['val_loss'], label='Denoising Validation Loss')
plt.title('Denoising Training and Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()

plt.show()

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20515/20515 [=====] - 8s 8ms/step
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2642188/2642188 [=====] - 2s 8ms/step
Downloading data from https://storage.googleapis.com/tensorflow/tf-datasets/10b-labels.idx-ubyte.gz
1348134/1348134 [=====] - 8s 8ms/step
Downloading data from https://storage.googleapis.com/tensorflow/tf-datasets/10b-images.idx-ubyte.gz
442182/442182 [=====] - 1s 8ms/step
Epoch 1/100
235/235 [=====] - 9s 12ms/step - loss: 0.3606 - val_loss: 0.3119
Epoch 2/100
235/235 [=====] - 1s 4ms/step - loss: 0.3630 - val_loss: 0.3020
Epoch 3/100
235/235 [=====] - 1s 5ms/step - loss: 0.2947 - val_loss: 0.2940
Epoch 4/100
235/235 [=====] - 1s 6ms/step - loss: 0.2982 - val_loss: 0.2909
```

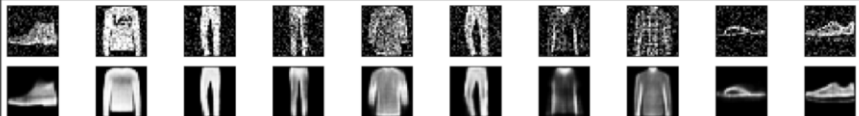
```
Untitled0.ipynb - Colab
colab.research.google.com/drive/1zo1Ntnb1z3fhS8KNLAp0vUf3rnnfr8gT?usp=chrome_ntp

Epoch 5/100
235/235 [=====] - 1s 5ms/step - loss: 0.2768 - val_loss: 0.2782
Epoch 6/100
235/235 [=====] - 1s 5ms/step - loss: 0.2758 - val_loss: 0.2781
Epoch 7/100
235/235 [=====] - 1s 5ms/step - loss: 0.2755 - val_loss: 0.2778
Epoch 8/100
235/235 [=====] - 1s 5ms/step - loss: 0.2752 - val_loss: 0.2774
Epoch 9/100
235/235 [=====] - 1s 5ms/step - loss: 0.2751 - val_loss: 0.2772
Epoch 10/100
235/235 [=====] - 1s 4ms/step - loss: 0.2749 - val_loss: 0.2771
Epoch 11/100
235/235 [=====] - 1s 4ms/step - loss: 0.2747 - val_loss: 0.2769
Epoch 12/100
235/235 [=====] - 1s 5ms/step - loss: 0.2745 - val_loss: 0.2768
Epoch 13/100
235/235 [=====] - 1s 6ms/step - loss: 0.2743 - val_loss: 0.2765
Epoch 14/100
235/235 [=====] - 1s 5ms/step - loss: 0.2742 - val_loss: 0.2768
Epoch 15/100
235/235 [=====] - 1s 5ms/step - loss: 0.2740 - val_loss: 0.2764
Epoch 16/100
235/235 [=====] - 1s 4ms/step - loss: 0.2739 - val_loss: 0.2763
Epoch 17/100
235/235 [=====] - 1s 4ms/step - loss: 0.2737 - val_loss: 0.2764
Epoch 18/100
235/235 [=====] - 1s 5ms/step - loss: 0.2735 - val_loss: 0.2758
Epoch 19/100
235/235 [=====] - 1s 4ms/step - loss: 0.2734 - val_loss: 0.2758
Epoch 20/100
235/235 [=====] - 1s 4ms/step - loss: 0.2733 - val_loss: 0.2757
Epoch 21/100
235/235 [=====] - 1s 4ms/step - loss: 0.2732 - val_loss: 0.2758
Epoch 22/100
235/235 [=====] - 1s 5ms/step - loss: 0.2730 - val_loss: 0.2753
Epoch 23/100
235/235 [=====] - 1s 4ms/step - loss: 0.2729 - val_loss: 0.2753
Epoch 24/100
235/235 [=====] - 1s 6ms/step - loss: 0.2728 - val_loss: 0.2752
Epoch 25/100
235/235 [=====] - 1s 6ms/step - loss: 0.2727 - val_loss: 0.2750
Epoch 26/100
235/235 [=====] - 1s 5ms/step - loss: 0.2726 - val_loss: 0.2754
Epoch 27/100
235/235 [=====] - 1s 4ms/step - loss: 0.2725 - val_loss: 0.2748
Epoch 28/100
235/235 [=====] - 1s 4ms/step - loss: 0.2724 - val_loss: 0.2750
Epoch 29/100
235/235 [=====] - 1s 4ms/step - loss: 0.2724 - val_loss: 0.2749
Epoch 30/100
235/235 [=====] - 1s 5ms/step - loss: 0.2722 - val_loss: 0.2747
Epoch 31/100
235/235 [=====] - 1s 5ms/step - loss: 0.2721 - val_loss: 0.2746
Epoch 32/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 33/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 34/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 35/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 36/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 37/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 38/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 39/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 40/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 41/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 42/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 43/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 44/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 45/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 46/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
Epoch 47/100
235/235 [=====] - 1s 4ms/step - loss: 0.2720 - val_loss: 0.2747
```

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Epoch 43/100 [=====] - 1s 4ms/step - loss: 0.2781 - val_loss: 0.2811
Epoch 44/100 [=====] - 1s 5ms/step - loss: 0.2780 - val_loss: 0.2810
Epoch 45/100 [=====] - 1s 6ms/step - loss: 0.2779 - val_loss: 0.2810
Epoch 46/100 [=====] - 1s 5ms/step - loss: 0.2779 - val_loss: 0.2812
Epoch 47/100 [=====] - 1s 5ms/step - loss: 0.2778 - val_loss: 0.2809
Epoch 48/100 [=====] - 1s 4ms/step - loss: 0.2777 - val_loss: 0.2808
Epoch 49/100 [=====] - 1s 4ms/step - loss: 0.2776 - val_loss: 0.2807
Epoch 50/100 [=====] - 1s 4ms/step - loss: 0.2776 - val_loss: 0.2807
Epoch 51/100 [=====] - 1s 4ms/step - loss: 0.2776 - val_loss: 0.2807
Epoch 52/100 [=====] - 1s 4ms/step - loss: 0.2775 - val_loss: 0.2806
Epoch 53/100 [=====] - 1s 5ms/step - loss: 0.2774 - val_loss: 0.2807
Epoch 54/100 [=====] - 1s 4ms/step - loss: 0.2774 - val_loss: 0.2806
Epoch 55/100 [=====] - 1s 5ms/step - loss: 0.2774 - val_loss: 0.2807
Epoch 56/100 [=====] - 1s 4ms/step - loss: 0.2774 - val_loss: 0.2806
Epoch 57/100 [=====] - 1s 4ms/step - loss: 0.2773 - val_loss: 0.2804
Epoch 58/100 [=====] - 1s 6ms/step - loss: 0.2773 - val_loss: 0.2805
Epoch 59/100 [=====] - 1s 7ms/step - loss: 0.2772 - val_loss: 0.2806
Epoch 60/100 [=====] - 1s 5ms/step - loss: 0.2772 - val_loss: 0.2806
Epoch 61/100 [=====] - 1s 5ms/step - loss: 0.2772 - val_loss: 0.2803
Epoch 62/100 [=====] - 1s 4ms/step - loss: 0.2771 - val_loss: 0.2804
Epoch 63/100 [=====] - 1s 4ms/step - loss: 0.2771 - val_loss: 0.2804
Epoch 64/100 [=====] - 1s 5ms/step - loss: 0.2770 - val_loss: 0.2804
Epoch 65/100 [=====] - 1s 5ms/step - loss: 0.2770 - val_loss: 0.2804
Epoch 66/100 [=====] - 1s 5ms/step - loss: 0.2769 - val_loss: 0.2804
Epoch 67/100 [=====] - 1s 5ms/step - loss: 0.2769 - val_loss: 0.2803
Epoch 68/100 [=====] - 1s 4ms/step - loss: 0.2769 - val_loss: 0.2802
Epoch 69/100 [=====] - 1s 6ms/step - loss: 0.2769 - val_loss: 0.2807
Epoch 70/100 [=====] - 1s 6ms/step - loss: 0.2768 - val_loss: 0.2801
Epoch 71/100 [=====] - 1s 5ms/step - loss: 0.2768 - val_loss: 0.2800
Epoch 72/100 [=====] - 1s 5ms/step - loss: 0.2768 - val_loss: 0.2800
Epoch 73/100 [=====] - 1s 5ms/step - loss: 0.2767 - val_loss: 0.2800
Epoch 74/100 [=====] - 1s 5ms/step - loss: 0.2767 - val_loss: 0.2800
```

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Epoch 81/100 [=====] - 1s 5ms/step - loss: 0.2766 - val_loss: 0.2798
Epoch 82/100 [=====] - 1s 6ms/step - loss: 0.2764 - val_loss: 0.2799
Epoch 83/100 [=====] - 1s 7ms/step - loss: 0.2764 - val_loss: 0.2797
Epoch 84/100 [=====] - 1s 5ms/step - loss: 0.2764 - val_loss: 0.2797
Epoch 85/100 [=====] - 1s 4ms/step - loss: 0.2764 - val_loss: 0.2798
Epoch 86/100 [=====] - 1s 5ms/step - loss: 0.2764 - val_loss: 0.2798
Epoch 87/100 [=====] - 1s 5ms/step - loss: 0.2763 - val_loss: 0.2796
Epoch 88/100 [=====] - 1s 5ms/step - loss: 0.2763 - val_loss: 0.2797
Epoch 89/100 [=====] - 1s 5ms/step - loss: 0.2763 - val_loss: 0.2796
Epoch 90/100 [=====] - 1s 4ms/step - loss: 0.2762 - val_loss: 0.2795
Epoch 91/100 [=====] - 1s 5ms/step - loss: 0.2762 - val_loss: 0.2796
Epoch 92/100 [=====] - 1s 5ms/step - loss: 0.2762 - val_loss: 0.2795
Epoch 93/100 [=====] - 1s 6ms/step - loss: 0.2762 - val_loss: 0.2794
Epoch 94/100 [=====] - 1s 6ms/step - loss: 0.2763 - val_loss: 0.2797
Epoch 95/100 [=====] - 1s 4ms/step - loss: 0.2763 - val_loss: 0.2798
Epoch 96/100 [=====] - 1s 4ms/step - loss: 0.2763 - val_loss: 0.2794
Epoch 97/100 [=====] - 1s 5ms/step - loss: 0.2763 - val_loss: 0.2794
Epoch 98/100 [=====] - 1s 4ms/step - loss: 0.2763 - val_loss: 0.2794
Epoch 99/100 [=====] - 1s 4ms/step - loss: 0.2760 - val_loss: 0.2795
Epoch 100/100 [=====] - 1s 5ms/step - loss: 0.2760 - val_loss: 0.2794
Epoch 100/100 [=====] - 1s 4ms/step - loss: 0.2760 - val_loss: 0.2795
113/113 [=====] - 1s 2ms/step
```



Training and Validation Loss

