



Experiment No.5

Aim: Design the architecture and implement the autoencoder model for Image Compression.

Code:

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

# Load and preprocess data
(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:])))

print(x_train.shape)
print(x_test.shape)

# Define the size of the encoded representation
encoding_dim = 32

# Define the model
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))

# Compile the model
autoencoder.compile(optimizer='adam', loss='binary_crossentropy')
```



```
# Train the model
autoencoder.fit(x_train, x_train,
                epochs=50,
                batch_size=256,
                shuffle=True,
                validation_data=(x_test, x_test))
```

Output:

Downloading data from

<https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz>

11490434/11490434 ————— 0s 0us/step

(60000, 784)

(10000, 784)

Epoch 1/50

235/235 ————— 4s 11ms/step - loss: 0.3849 -

val_loss: 0.1924

Epoch 2/50

235/235 ————— 5s 9ms/step - loss: 0.1829 -

val_loss: 0.1554

Epoch 3/50

235/235 ————— 3s 10ms/step - loss: 0.1508 -

val_loss: 0.1343

Epoch 4/50

235/235 ————— 4s 15ms/step - loss: 0.1317 -

val_loss: 0.1198

Epoch 5/50

235/235 ————— 5s 15ms/step - loss: 0.1190 -

val_loss: 0.1114

Epoch 6/50

235/235 ————— 4s 9ms/step - loss: 0.1112 -



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val_loss: 0.1059

Epoch 7/50

235/235 ————— 4s 17ms/step - loss: 0.1058 -

val_loss: 0.1017

Epoch 8/50

235/235 ————— 2s 9ms/step - loss: 0.1021 -

val_loss: 0.0986

Epoch 9/50

235/235 ————— 3s 9ms/step - loss: 0.0993 -

val_loss: 0.0965

Epoch 10/50

235/235 ————— 3s 10ms/step - loss: 0.0975 -

val_loss: 0.0950

Epoch 11/50

235/235 ————— 3s 11ms/step - loss: 0.0964 -

val_loss: 0.0942

Epoch 12/50

235/235 ————— 4s 16ms/step - loss: 0.0951 -

val_loss: 0.0934

Epoch 13/50

235/235 ————— 4s 10ms/step - loss: 0.0948 -

val_loss: 0.0930

Epoch 14/50

235/235 ————— 2s 10ms/step - loss: 0.0944 -

val_loss: 0.0928

Epoch 15/50

235/235 ————— 3s 10ms/step - loss: 0.0940 -

val_loss: 0.0926

Epoch 16/50



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235/235 ————— 3s 12ms/step - loss: 0.0938 -

val_loss: 0.0924

Epoch 17/50

235/235 ————— 3s 12ms/step - loss: 0.0936 -

val_loss: 0.0923

Epoch 18/50

235/235 ————— 2s 9ms/step - loss: 0.0933 -

val_loss: 0.0922

Epoch 19/50

235/235 ————— 3s 10ms/step - loss: 0.0933 -

val_loss: 0.0920

Epoch 20/50

235/235 ————— 2s 9ms/step - loss: 0.0932 -

val_loss: 0.0920

Epoch 21/50

235/235 ————— 3s 11ms/step - loss: 0.0932 -

val_loss: 0.0920

Epoch 22/50

235/235 ————— 3s 14ms/step - loss: 0.0931 -

val_loss: 0.0920

Epoch 23/50

235/235 ————— 2s 9ms/step - loss: 0.0932 -

val_loss: 0.0919

Epoch 24/50

235/235 ————— 3s 9ms/step - loss: 0.0932 -

val_loss: 0.0918

Epoch 25/50

235/235 ————— 3s 9ms/step - loss: 0.0932 -



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val_loss: 0.0918

Epoch 26/50

235/235 ————— 3s 11ms/step - loss: 0.0930 -

val_loss: 0.0918

Epoch 27/50

235/235 ————— 5s 9ms/step - loss: 0.0929 -

val_loss: 0.0917

Epoch 28/50

235/235 ————— 3s 10ms/step - loss: 0.0928 -

val_loss: 0.0917

Epoch 29/50

235/235 ————— 2s 9ms/step - loss: 0.0929 -

val_loss: 0.0917

Epoch 30/50

235/235 ————— 3s 12ms/step - loss: 0.0927 -

val_loss: 0.0916

Epoch 31/50

235/235 ————— 5s 10ms/step - loss: 0.0927 -

val_loss: 0.0916

Epoch 32/50

235/235 ————— 3s 10ms/step - loss: 0.0927 -

val_loss: 0.0916

Epoch 33/50

235/235 ————— 2s 9ms/step - loss: 0.0928 -

val_loss: 0.0916

Epoch 34/50

235/235 ————— 3s 12ms/step - loss: 0.0926 -

val_loss: 0.0916

Epoch 35/50



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235/235 ————— 3s 12ms/step - loss: 0.0926 -

val_loss: 0.0915

Epoch 36/50

235/235 ————— 2s 9ms/step - loss: 0.0928 -

val_loss: 0.0916

Epoch 37/50

235/235 ————— 3s 9ms/step - loss: 0.0929 -

val_loss: 0.0915

Epoch 38/50

235/235 ————— 2s 9ms/step - loss: 0.0929 -

val_loss: 0.0915

Epoch 39/50

235/235 ————— 3s 11ms/step - loss: 0.0925 -

val_loss: 0.0915

Epoch 40/50

235/235 ————— 5s 10ms/step - loss: 0.0928 -

val_loss: 0.0915

Epoch 41/50

235/235 ————— 3s 10ms/step - loss: 0.0927 -

val_loss: 0.0915

Epoch 42/50

235/235 ————— 2s 9ms/step - loss: 0.0929 -

val_loss: 0.0915

Epoch 43/50

235/235 ————— 3s 12ms/step - loss: 0.0926 -

val_loss: 0.0915

Epoch 44/50

235/235 ————— 4s 9ms/step - loss: 0.0926 -



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val_loss: 0.0915

Epoch 45/50

235/235 ————— 3s 9ms/step - loss: 0.0925 -

val_loss: 0.0915

Epoch 46/50

235/235 ————— 3s 9ms/step - loss: 0.0927 -

val_loss: 0.0915

Epoch 47/50

235/235 ————— 3s 13ms/step - loss: 0.0925 -

val_loss: 0.0915

Epoch 48/50

235/235 ————— 4s 9ms/step - loss: 0.0924 -

val_loss: 0.0914

Epoch 49/50

235/235 ————— 3s 10ms/step - loss: 0.0928 -

val_loss: 0.0915

Epoch 50/50

235/235 ————— 2s 10ms/step - loss: 0.0925 -

val_loss: 0.0915

<keras.src.callbacks.history.History at 0x7c59d720fca0>

Code:

```
# Encode some digits from the test set
```

```
encoded_imgs = encoder.predict(x_test)
```

```
# Decode the encoded images
```

```
decoded_imgs = decoder.predict(encoded_imgs)
```



Output:

313/313 ————— 0s 1ms/step

313/313 ————— 2s 6ms/step

Code:

```
import matplotlib.pyplot as plt

n = 10 # Number of digits to 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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Output:

