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import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras.layers import Input, Dense
from tensorflow.keras.models import Model
from tensorflow.keras.datasets import cifar10

# Load the CIFAR-10 dataset
(x_train, _), (x_test, _) = cifar10.load_data()

# Normalize pixel values to be between 0 and 1
x_train = x_train.astype('float32') / 255.0
x_test = x_test.astype('float32') / 255.0

# Flatten the images for the autoencoder
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 the autoencoder model
encoding_dim = 128 # Size of the encoded representations
input_img = Input(shape=(3072,)) # 32x32x3 images flattened
encoded = Dense(encoding_dim, activation='relu')(input_img)
decoded = Dense(3072, activation='sigmoid')(encoded)

autoencoder = Model(input_img, decoded)

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

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

# Create a separate encoder model
encoder = Model(input_img, encoded)

# Encode the test images
encoded_imgs = encoder.predict(x_test)

# Decode the encoded images
decoded_imgs = autoencoder.predict(x_test)

# Display original and reconstructed images
n = 10 # Number of images to display
plt.figure(figsize=(20, 4))
for i in range(n):
    # Original images
    ax = plt.subplot(2, n, i + 1)
    plt.imshow(x_test[i].reshape(32, 32, 3))
    plt.gray()
    ax.get_xaxis().set_visible(False)

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ax.get_yaxis().set_visible(False)

# Reconstructed images
ax = plt.subplot(2, n, i + 1 + n)
plt.imshow(decoded_imgs[i].reshape(32, 32, 3))
plt.gray()
ax.get_xaxis().set_visible(False)
ax.get_yaxis().set_visible(False)

plt.show()

Downloading data from https://www.cs.toronto.edu/~kriz/cifar-10-
python.tar.gz
170498071/170498071 [=====] - 4s 0us/step
Epoch 1/50
196/196 [=====] - 13s 58ms/step - loss:
0.6391 - val_loss: 0.6210
Epoch 2/50
196/196 [=====] - 10s 51ms/step - loss:
0.6133 - val_loss: 0.6086
Epoch 3/50
196/196 [=====] - 9s 46ms/step - loss: 0.6043
- val_loss: 0.6009
Epoch 4/50
196/196 [=====] - 9s 45ms/step - loss: 0.5984
- val_loss: 0.5963
Epoch 5/50
196/196 [=====] - 9s 48ms/step - loss: 0.5933
- val_loss: 0.5969
Epoch 6/50
196/196 [=====] - 10s 51ms/step - loss:
0.5906 - val_loss: 0.5899
Epoch 7/50
196/196 [=====] - 9s 46ms/step - loss: 0.5883
- val_loss: 0.5880
Epoch 8/50
196/196 [=====] - 9s 46ms/step - loss: 0.5869
- val_loss: 0.5888
Epoch 9/50
196/196 [=====] - 9s 47ms/step - loss: 0.5859
- val_loss: 0.5865
Epoch 10/50
196/196 [=====] - 9s 48ms/step - loss: 0.5858
- val_loss: 0.5860
Epoch 11/50
196/196 [=====] - 10s 51ms/step - loss:
0.5855 - val_loss: 0.5859
Epoch 12/50
196/196 [=====] - 9s 46ms/step - loss: 0.5848
- val_loss: 0.5862

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Epoch 13/50
196/196 [=====] - 9s 48ms/step - loss: 0.5845
- val_loss: 0.5859
Epoch 14/50
196/196 [=====] - 10s 51ms/step - loss:
0.5847 - val_loss: 0.5857
Epoch 15/50
196/196 [=====] - 9s 44ms/step - loss: 0.5846
- val_loss: 0.5850
Epoch 16/50
196/196 [=====] - 9s 47ms/step - loss: 0.5846
- val_loss: 0.5863
Epoch 17/50
196/196 [=====] - 9s 48ms/step - loss: 0.5843
- val_loss: 0.5850
Epoch 18/50
196/196 [=====] - 9s 48ms/step - loss: 0.5845
- val_loss: 0.5860
Epoch 19/50
196/196 [=====] - 8s 43ms/step - loss: 0.5843
- val_loss: 0.5850
Epoch 20/50
196/196 [=====] - 9s 48ms/step - loss: 0.5841
- val_loss: 0.5850
Epoch 21/50
196/196 [=====] - 9s 48ms/step - loss: 0.5843
- val_loss: 0.5852
Epoch 22/50
196/196 [=====] - 9s 47ms/step - loss: 0.5844
- val_loss: 0.5848
Epoch 23/50
196/196 [=====] - 9s 47ms/step - loss: 0.5840
- val_loss: 0.5854
Epoch 24/50
196/196 [=====] - 9s 48ms/step - loss: 0.5842
- val_loss: 0.5856
Epoch 25/50
196/196 [=====] - 9s 48ms/step - loss: 0.5843
- val_loss: 0.5852
Epoch 26/50
196/196 [=====] - 9s 46ms/step - loss: 0.5841
- val_loss: 0.5848
Epoch 27/50
196/196 [=====] - 9s 45ms/step - loss: 0.5843
- val_loss: 0.5855
Epoch 28/50
196/196 [=====] - 9s 48ms/step - loss: 0.5840
- val_loss: 0.5848
Epoch 29/50
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196/196 [=====] - 9s 48ms/step - loss: 0.5843
- val_loss: 0.5854
Epoch 30/50
196/196 [=====] - 9s 46ms/step - loss: 0.5839
- val_loss: 0.5848
Epoch 31/50
196/196 [=====] - 9s 46ms/step - loss: 0.5840
- val_loss: 0.5853
Epoch 32/50
196/196 [=====] - 9s 48ms/step - loss: 0.5842
- val_loss: 0.5849
Epoch 33/50
196/196 [=====] - 9s 48ms/step - loss: 0.5841
- val_loss: 0.5847
Epoch 34/50
196/196 [=====] - 8s 43ms/step - loss: 0.5839
- val_loss: 0.5850
Epoch 35/50
196/196 [=====] - 9s 48ms/step - loss: 0.5840
- val_loss: 0.5847
Epoch 36/50
196/196 [=====] - 9s 48ms/step - loss: 0.5843
- val_loss: 0.5851
Epoch 37/50
196/196 [=====] - 9s 46ms/step - loss: 0.5839
- val_loss: 0.5852
Epoch 38/50
196/196 [=====] - 10s 52ms/step - loss:
0.5839 - val_loss: 0.5852
Epoch 39/50
196/196 [=====] - 9s 48ms/step - loss: 0.5841
- val_loss: 0.5847
Epoch 40/50
196/196 [=====] - 9s 48ms/step - loss: 0.5840
- val_loss: 0.5853
Epoch 41/50
196/196 [=====] - 9s 48ms/step - loss: 0.5841
- val_loss: 0.5851
Epoch 42/50
196/196 [=====] - 8s 43ms/step - loss: 0.5839
- val_loss: 0.5846
Epoch 43/50
196/196 [=====] - 9s 48ms/step - loss: 0.5840
- val_loss: 0.5851
Epoch 44/50
196/196 [=====] - 9s 48ms/step - loss: 0.5842
- val_loss: 0.5857
Epoch 45/50
196/196 [=====] - 9s 47ms/step - loss: 0.5839
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- val_loss: 0.5848
Epoch 46/50
196/196 [=====] - 9s 44ms/step - loss: 0.5839
- val_loss: 0.5846
Epoch 47/50
196/196 [=====] - 9s 48ms/step - loss: 0.5839
- val_loss: 0.5847
Epoch 48/50
196/196 [=====] - 9s 48ms/step - loss: 0.5842
- val_loss: 0.5848
Epoch 49/50
196/196 [=====] - 9s 46ms/step - loss: 0.5839
- val_loss: 0.5854
Epoch 50/50
196/196 [=====] - 9s 46ms/step - loss: 0.5839
- val_loss: 0.5851
313/313 [=====] - 1s 3ms/step
313/313 [=====] - 1s 3ms/step
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