In [1]: from keras.datasets import cifar10 (x_train,_),(x_test,_)=cifar10.load_data() Downloading data from https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz import matplotlib.pyplot as plt In [7]: fig, axes = plt.subplots(ncols=7, nrows=3, figsize=(17, 8)) index = 0for i in range(3): for j in range(7): axes[i,j].imshow(x_train[index]) axes[i,j].get_xaxis().set_visible(False) axes[i,j].get_yaxis().set_visible(False) index += 1plt.show() import numpy as np In [11]: import cv2 x_train = np.array([cv2.cvtColor(image, cv2.COLOR_BGR2GRAY) for image in x_train]) x_test = np.array([cv2.cvtColor(image, cv2.COLOR_BGR2GRAY) for image in x_test]) ig, axes = plt.subplots(ncols=7, nrows=3, figsize=(17, 8)) In [12]: index = 0for i in range(3): for j in range(7): axes[i,j].imshow(x_train[index], cmap='gray') axes[i,j].get_xaxis().set_visible(False) axes[i,j].get_yaxis().set_visible(False) index += 1plt.show() In [13]: print(x_train.shape) (50000, 32, 32) In [14]: def pre_process(X): X = X/255.0X = X.reshape((len(X), 1024))return X x_train = pre_process(x_train) x_test = pre_process(x_test) print("X_train", x_train.shape) print("X_test", x_test.shape) X_train (50000, 1024) X_test (10000, 1024) import matplotlib.pyplot as plt In [15]: def show_data(X, n=10, height=32, width=32, title=""): plt.figure(figsize=(20, 5)) **for** i **in** range(n): ax = plt.subplot(2, n, i+1)plt.imshow(X[i].reshape((height,width))) plt.gray() ax.get_xaxis().set_visible(False) ax.get_yaxis().set_visible(False) plt.suptitle(title, fontsize = 20) show_data(x_train, title="Training images") Training images In [16]: from keras.layers import Input, Dense from keras.models import Model In [30]: # encoder input_layer = Input(shape=(1024,), name="INPUT") hidden_layer_1 = Dense(512, activation='relu', name="HIDDEN_1")(input_layer) hidden_layer_2 = Dense(256, activation='relu', name="HIDDEN_2")(hidden_layer_1) # code code_layer = Dense(150, activation='relu', name="CODE")(hidden_layer_2) # decoder hidden_layer_3 = Dense(256, activation='relu', name="HIDDEN_3")(code_layer) hidden_layer_4 = Dense(512, activation='relu', name="HIDDEN_4")(hidden_layer_3) output_layer = Dense(1024, activation='sigmoid', name="OUTPUT")(hidden_layer_4) In [31]: stacked_autoencoder=Model(input_layer, output_layer) stacked_autoencoder.compile(optimizer="Adam", loss="mse") stacked_autoencoder.summary() Model: "model_4" Layer (type) Output Shape Param # INPUT (InputLayer) [(None, 1024)] HIDDEN_1 (Dense) (None, 512) 524800 HIDDEN_2 (Dense) 131328 (None, 256) CODE (Dense) (None, 150) 38550 HIDDEN_3 (Dense) (None, 256) 38656 HIDDEN_4 (Dense) (None, 512) 131584 OUTPUT (Dense) (None, 1024) 525312 Total params: 1,390,230 Trainable params: 1,390,230 Non-trainable params: 0 In [32]: stacked_autoencoder.fit(x_train, x_train, epochs=100,batch_size=256,shuffle=True,validation_data=(x_test, x_ Epoch 1/100 Epoch 2/100 Epoch 3/100 Epoch 4/100 Epoch 5/100 Epoch 6/100 196/196 [==== Epoch 7/100 196/196 [===== Epoch 8/100 Epoch 9/100 Epoch 10/100 Epoch 11/100 196/196 [====== Epoch 12/100 196/196 [====== Epoch 13/100 196/196 [====== Epoch 14/100 196/196 [====== Epoch 15/100 196/196 [====== Epoch 16/100 Epoch 17/100 Epoch 18/100 ===============] - 10s 53ms/step - loss: 0.0086 - val_loss: 0.0089 196/196 [====== Epoch 19/100 196/196 [===== Epoch 20/100 ========] - 10s 53ms/step - loss: 0.0083 - val_loss: 0.0090 196/196 [===== Epoch 21/100 196/196 [====== Epoch 22/100 ===========] - 11s 55ms/step - loss: 0.0081 - val_loss: 0.0080 196/196 [===== Epoch 23/100 196/196 [===== ============] - 11s 55ms/step - loss: 0.0080 - val_loss: 0.0078 Epoch 24/100 196/196 [====== Epoch 25/100 Epoch 26/100 196/196 [====== Epoch 27/100 196/196 [====== Epoch 28/100 Epoch 29/100 Epoch 30/100 Epoch 31/100 10s 53ms/step - loss: 0.0072 - val_loss: 0.0077 196/196 [= Epoch 32/100 Epoch 33/100 Epoch 34/100 Epoch 35/100 Epoch 36/100 Epoch 37/100 Epoch 38/100 Epoch 39/100 Epoch 40/100 Epoch 41/100 Epoch 42/100 Epoch 43/100 Epoch 44/100 Epoch 45/100 Epoch 46/100 Epoch 47/100 Epoch 48/100 Epoch 49/100 Epoch 50/100 Epoch 51/100 Epoch 52/100 Epoch 53/100 Epoch 54/100 Epoch 55/100 Epoch 56/100 Epoch 57/100 Epoch 58/100 Epoch 59/100 Epoch 60/100 Epoch 61/100 Epoch 62/100 Epoch 63/100 Epoch 64/100 Epoch 65/100 Epoch 66/100 Epoch 67/100 Epoch 68/100 Epoch 69/100 Epoch 70/100 Epoch 71/100 =======] - 11s 55ms/step - loss: 0.0060 - val_loss: 0.0060 196/196 [= Epoch 72/100 196/196 [======= Epoch 73/100 Epoch 74/100 196/196 [===== Epoch 75/100 196/196 [==== Epoch 76/100 196/196 [===== Epoch 77/100 196/196 [====== Epoch 78/100 196/196 [====== Epoch 79/100 196/196 [===== ========] - 11s 55ms/step - loss: 0.0059 - val_loss: 0.0058 Epoch 80/100 196/196 [===== Epoch 81/100 Epoch 82/100 Epoch 83/100 196/196 [====== Epoch 84/100 196/196 [==== Epoch 85/100 Epoch 86/100 196/196 [====== Epoch 87/100 196/196 [==== Epoch 88/100 196/196 [===== Epoch 89/100 Epoch 90/100 196/196 [====== Epoch 91/100 196/196 [==== Epoch 92/100 196/196 [===== Epoch 93/100 =========] - 11s 56ms/step - loss: 0.0056 - val_loss: 0.0056 196/196 [==== Epoch 94/100 =======] - 11s 55ms/step - loss: 0.0056 - val_loss: 0.0056 196/196 [==== Epoch 95/100 196/196 [==== ========] - 11s 55ms/step - loss: 0.0056 - val_loss: 0.0056 Epoch 96/100 ========] - 11s 56ms/step - loss: 0.0056 - val_loss: 0.0056 196/196 [==== Epoch 97/100 ========] - 11s 55ms/step - loss: 0.0055 - val_loss: 0.0057 196/196 [==== Epoch 98/100 ==========] - 11s 55ms/step - loss: 0.0056 - val_loss: 0.0056 196/196 [==== Epoch 99/100 196/196 [===== ==========] - 11s 55ms/step - loss: 0.0056 - val_loss: 0.0055 Epoch 100/100 196/196 [===== Out[32]: <tensorflow.python.keras.callbacks.History at 0x7fc2fd037828> decoded_data = stacked_autoencoder.predict(x_test) In [33]: In [34]: show_data(x_test, title="original data") show_data(decoded_data, title="decoded data") original data decoded data In []:

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