Activity #5

Image Enhancement with

AutoEncoder



Agenda

5.1 Data Preparation

5.2 Adding Noise

5.3 Autoencoder



Libraries

- import numpy as np
 import cv2
 from matplotlib import pyplot as plt
 - from keras.models import Model, Input
- from tensorflow.keras.layers import Dense, Conv2D, MaxPool2D, UpSampling2D
 - from tensorflow.keras.callbacks import EarlyStopping
 - from keras.preprocessing import image
 - from sklearn.model_selection import train_test_split
 - Import glob



DATA PREPARATION

5.1 Data Preparation

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- # List all filename in face image path
 - filenames= glob.glob ()

- # Load image files and Resize (h,w,ch) -> h = w < 100, ch = 3 (R,G,B)
 - img = image.load_img(fname, target_size, interpolation="nearest")
- # convert the image to an array
 - img = img.img_to_array(img)
- # Normalized color image
 - img = img/255
- all_imgs = img.append()

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- # convert to numpy array
- all_images = np.array(all_images)

- # split data into train and validation data
- train_x, val_x = train_test_split(all_images, random_state=32, test_size=0.3)



ADD NOISE

4.2 Prepare input image from scratch

• # Add Noise

• กำหนด noise factor (noise level)

• # Add noise

• noise_factor = scalar between 0 -1 ลองอย่างน้อย 2 ค่า

• # กำหนด noise parameter

Noise distribution: normal

• Noise mean: zero mean (Nmean = 0)

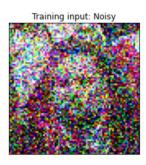
Noise std: unit variance (Nstd = 1)

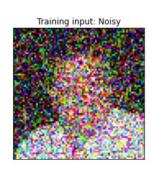
• # use np.random.normal to generate normal distribution (gaussian) noise

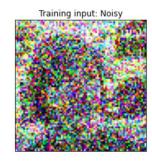
• x_train_noisy = x_train + (noise_factor * np.random.normal(loc=Nmean, scale=Nstd, size=x_train.shape))

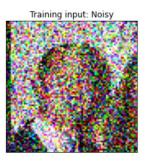
• x_val_noisy = x_val + (noise_factor * np.random.normal(loc=Nmean, scale=Nstd, size=x_val.shape))

Training input: Noisy









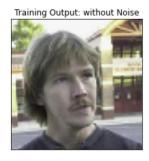
Training Output: without Noise



Training Output: without Noise









Adding Noise

Choose to plot at least 3 images



AUTOENCODER MODEL

5.3 Autoencoder Architecture

- # กำหนด Encoder Architecture
 - Input_img = Input(shape=(height, width, ch))
 - #encoding architecture
 - x1 = Conv2D(256, (3, 3), activation='relu', padding='same')(Input_img)
 - x2 = Conv2D(128, (3, 3), activation='relu', padding='same')(x1)
- x2 = MaxPool2D((2, 2))(x2)
- encoded = Conv2D(64, (3, 3), activation='relu', padding='same')(x2)
- # กำหนด Decoder Architecture
- # decoding architecture
- x3 = Conv2D(64, (3, 3), activation='relu', padding='same')(encoded)
- x3 = UpSampling2D((2, 2))(x3)
- x2 = Conv2D(128, (3, 3), activation='relu', padding='same')(x3)
- x1 = Conv2D(256, (3, 3), activation='relu', padding='same')(x2)
- decoded = Conv2D(3, (3, 3), padding='same')(x1)
- # กำหนด optimizer setting
- autoencoder = Model(Input_img, decoded)
- autoencoder.compile(optimizer='adam', loss='mse') -> loss ใช้ Mean Square Error
- autoencoder.summary()

Model: "model"

Layer (type)	Output Shape	Param #
input_2 (InputLayer)	[(None, 80, 80, 3)]	0
conv2d (Conv2D)	(None, 80, 80, 256)	7168
conv2d_1 (Conv2D)	(None, 80, 80, 128)	295040
<pre>max_pooling2d (MaxPooling2D)</pre>	(None, 40, 40, 128)	0
conv2d_2 (Conv2D)	(None, 40, 40, 64)	73792
conv2d_3 (Conv2D)	(None, 40, 40, 64)	36928
<pre>up_sampling2d (UpSampling2D)</pre>	(None, 80, 80, 64)	0
conv2d_4 (Conv2D)	(None, 80, 80, 128)	73856
conv2d_5 (Conv2D)	(None, 80, 80, 256)	295168
conv2d_6 (Conv2D)	(None, 80, 80, 3)	6915

Total params: 788,867

Trainable params: 788,867 Non-trainable params: 0

5.2

Autoencoder Model

5.3 Training Autoencoder Model

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- # กำหนด Training parameter
 - epoch ทดลองอย่างน้อย 2 ค่า [2, 3, 4]
- batch_size ทดลองอย่างน้อย 1 ค่า [8, 16, 32]
- # เริ่มการ training
- history = autoencoder.fit (x_train_noisy, train_x,
- epochs=2,
- batch size=32,
- shuffle=True,
- validation_data=(x_val_noisy, val_x),
- callbacks=[early_stopper])

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- # ทุคสอบ autoencoder model ด้วย validation
- predictions = autoencoder.predict(x_val_noisy)
- View Loss from history
 - plt.plot(history.history['loss'])
 - plt.plot(history.history['val_loss'])
- plt.title('model loss') plt.ylabel('loss')
- plt.xlabel('epoch')
- plt.legend(['train', 'test'], loc='upper left')
- plt.show()
- # Display Result image

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Autoencoder Model

