

DL EXP 6 - Design the Architecture & Implement the Autoencoder Model for Image Compression.

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DL EXPERIMENT NO : 06

```
import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Input, Conv2D, MaxPooling2D, UpSampling2D
from PIL import Image
```

```
image_path = '/content/Paree.jpg'
img = Image.open(image_path)
img = img.resize((128, 128))
img = img.convert('L')
img_array = np.asarray(img) / 255.0
img_array = np.reshape(img_array, (1, 128, 128, 1))
```

```
input_img = Input(shape=(128, 128, 1))
```

```
# Encoder
x = Conv2D(32, (3, 3), activation='relu', padding='same')(input_img)
x = MaxPooling2D((2, 2), padding='same')(x)
x = Conv2D(16, (3, 3), activation='relu', padding='same')(x)
x = MaxPooling2D((2, 2), padding='same')(x)
x = Conv2D(8, (3, 3), activation='relu', padding='same')(x)
encoded = MaxPooling2D((2, 2), padding='same')(x)
```

```
# Decoder
x = Conv2D(8, (3, 3), activation='relu', padding='same')(encoded)
x = UpSampling2D((2, 2))(x)
x = Conv2D(16, (3, 3), activation='relu', padding='same')(x)
x = UpSampling2D((2, 2))(x)
x = Conv2D(32, (3, 3), activation='relu', padding='same')(x)
x = UpSampling2D((2, 2))(x)
decoded = Conv2D(1, (3, 3), activation='sigmoid', padding='same')(x)
```

```
# Autoencoder Model
autoencoder = Model(input_img, decoded)
autoencoder.compile(optimizer='adam', loss='binary_crossentropy')
```

```
autoencoder.summary()
```

Model: "functional"

Layer (type)	Output Shape	Param #
input_layer (InputLayer)	(None, 128, 128, 1)	0
conv2d (Conv2D)	(None, 128, 128, 32)	320
max_pooling2d (MaxPooling2D)	(None, 64, 64, 32)	0
conv2d_1 (Conv2D)	(None, 64, 64, 16)	4,624
max_pooling2d_1 (MaxPooling2D)	(None, 32, 32, 16)	0
conv2d_2 (Conv2D)	(None, 32, 32, 8)	1,160
max_pooling2d_2 (MaxPooling2D)	(None, 16, 16, 8)	0
conv2d_3 (Conv2D)	(None, 16, 16, 8)	584
up_sampling2d (UpSampling2D)	(None, 32, 32, 8)	0
conv2d_4 (Conv2D)	(None, 32, 32, 16)	1,168
up_sampling2d_1 (UpSampling2D)	(None, 64, 64, 16)	0
conv2d_5 (Conv2D)	(None, 64, 64, 32)	4,640
up_sampling2d_2 (UpSampling2D)	(None, 128, 128, 32)	0
conv2d_6 (Conv2D)	(None, 128, 128, 1)	289

Total params: 12,785 (49.94 KB)
Trainable params: 12,785 (49.94 KB)

```
autoencoder.fit(img_array, img_array,
                epochs=100,
                batch_size=1,
                shuffle=True)
```

```
Epoch 1/100
1/1 ————— 3s 3s/step - loss: 0.6976
Epoch 2/100
1/1 ————— 0s 457ms/step - loss: 0.6940
Epoch 3/100
1/1 ————— 0s 271ms/step - loss: 0.6928
Epoch 4/100
1/1 ————— 0s 148ms/step - loss: 0.6921
Epoch 5/100
1/1 ————— 0s 142ms/step - loss: 0.6916
Epoch 6/100
1/1 ————— 0s 129ms/step - loss: 0.6910
Epoch 7/100
1/1 ————— 0s 143ms/step - loss: 0.6902
Epoch 8/100
1/1 ————— 0s 85ms/step - loss: 0.6894
Epoch 9/100
1/1 ————— 0s 133ms/step - loss: 0.6884
Epoch 10/100
1/1 ————— 0s 78ms/step - loss: 0.6870
Epoch 11/100
1/1 ————— 0s 77ms/step - loss: 0.6853
Epoch 12/100
1/1 ————— 0s 142ms/step - loss: 0.6833
Epoch 13/100
1/1 ————— 0s 132ms/step - loss: 0.6808
Epoch 14/100
1/1 ————— 0s 77ms/step - loss: 0.6779
Epoch 15/100
1/1 ————— 0s 90ms/step - loss: 0.6747
Epoch 16/100
1/1 ————— 0s 129ms/step - loss: 0.6718
Epoch 17/100
1/1 ————— 0s 77ms/step - loss: 0.6701
Epoch 18/100
1/1 ————— 0s 79ms/step - loss: 0.6709
Epoch 19/100
1/1 ————— 0s 77ms/step - loss: 0.6729
Epoch 20/100
1/1 ————— 0s 76ms/step - loss: 0.6728
Epoch 21/100
1/1 ————— 0s 144ms/step - loss: 0.6706
Epoch 22/100
1/1 ————— 0s 83ms/step - loss: 0.6680
Epoch 23/100
1/1 ————— 0s 74ms/step - loss: 0.6668
Epoch 24/100
1/1 ————— 0s 78ms/step - loss: 0.6672
Epoch 25/100
1/1 ————— 0s 142ms/step - loss: 0.6677
Epoch 26/100
1/1 ————— 0s 135ms/step - loss: 0.6677
Epoch 27/100
1/1 ————— 0s 73ms/step - loss: 0.6669
Epoch 28/100
1/1 ————— 0s 76ms/step - loss: 0.6656
Epoch 29/100
1/1 ————— 0s 136ms/step - loss: 0.6643
```

```
decoded_img = autoencoder.predict(img_array)
decoded_img = decoded_img.reshape(128, 128)
plt.figure(figsize=(10, 4))
```

```
plt.subplot(1, 2, 1)
plt.imshow(img_array.reshape(128, 128), cmap='gray')
plt.title("Original Image")
plt.axis('off')
```

```
plt.subplot(1, 2, 2)
plt.imshow(decoded_img, cmap='gray')
plt.title("Reconstructed Image")
plt.axis('off')
```

```
plt.show()
```

1/1 0s 399ms/step

Original Image



Reconstructed Image

