

```

import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras.layers import Input, Dense
from tensorflow.keras.models import Model
from sklearn.datasets import fetch_lfw_people
from sklearn.model_selection import train_test_split

# Load the LFW dataset (Labeled Faces in the Wild)
lfw_people = fetch_lfw_people(min_faces_per_person=70, resize=0.4)

# Split the dataset into training and testing sets
X_train, X_test, _, _ = train_test_split(lfw_people.data,
    lfw_people.target, test_size=0.25, random_state=42)

# 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

# Define the autoencoder model
encoding_dim = 64 # Size of the encoded representations
input_img = Input(shape=(X_train.shape[1],))
encoded = Dense(encoding_dim, activation='relu')(input_img)
decoded = Dense(X_train.shape[1], 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, 8)) # Adjusted figsize for two rows
for i in range(n):
    # Original images
    ax = plt.subplot(2, n, i + 1)
    plt.imshow(X_test[i].reshape(50, 37), cmap='gray') # Assuming
    images are 50x37 grayscale

```

```
ax.get_xaxis().set_visible(False)
ax.get_yaxis().set_visible(False)
```

```
plt.show()
```

```
Epoch 1/50
```

```
4/4 [=====] - 1s 76ms/step - loss: 0.6924 -  
val_loss: 0.6910
```

```
Epoch 2/50
```

```
4/4 [=====] - 0s 37ms/step - loss: 0.6901 -  
val_loss: 0.6882
```

```
Epoch 3/50
```

```
4/4 [=====] - 0s 34ms/step - loss: 0.6869 -  
val_loss: 0.6842
```

```
Epoch 4/50
```

```
4/4 [=====] - 0s 35ms/step - loss: 0.6823 -  
val_loss: 0.6783
```

```
Epoch 5/50
```

```
4/4 [=====] - 0s 36ms/step - loss: 0.6755 -  
val_loss: 0.6700
```

```
Epoch 6/50
```

```
4/4 [=====] - 0s 34ms/step - loss: 0.6661 -  
val_loss: 0.6587
```

```
Epoch 7/50
```

```
4/4 [=====] - 0s 35ms/step - loss: 0.6536 -  
val_loss: 0.6442
```

```
Epoch 8/50
```

```
4/4 [=====] - 0s 34ms/step - loss: 0.6378 -  
val_loss: 0.6263
```

```
Epoch 9/50
```

```
4/4 [=====] - 0s 32ms/step - loss: 0.6186 -  
val_loss: 0.6049
```

```
Epoch 10/50
```

```
4/4 [=====] - 0s 36ms/step - loss: 0.5957 -  
val_loss: 0.5800
```

```
Epoch 11/50
```

```
4/4 [=====] - 0s 32ms/step - loss: 0.5696 -  
val_loss: 0.5520
```

```
Epoch 12/50
```

```
4/4 [=====] - 0s 35ms/step - loss: 0.5403 -  
val_loss: 0.5212
```

```
Epoch 13/50
```

```
4/4 [=====] - 0s 32ms/step - loss: 0.5084 -  
val_loss: 0.4880
```

```
Epoch 14/50
```

```
4/4 [=====] - 0s 33ms/step - loss: 0.4744 -  
val_loss: 0.4531
```

```
Epoch 15/50
```

```
4/4 [=====] - 0s 35ms/step - loss: 0.4389 -  
val_loss: 0.4172  
Epoch 16/50  
4/4 [=====] - 0s 36ms/step - loss: 0.4027 -  
val_loss: 0.3810  
Epoch 17/50  
4/4 [=====] - 0s 35ms/step - loss: 0.3664 -  
val_loss: 0.3452  
Epoch 18/50  
4/4 [=====] - 0s 36ms/step - loss: 0.3309 -  
val_loss: 0.3107  
Epoch 19/50  
4/4 [=====] - 0s 33ms/step - loss: 0.2969 -  
val_loss: 0.2779  
Epoch 20/50  
4/4 [=====] - 0s 36ms/step - loss: 0.2649 -  
val_loss: 0.2474  
Epoch 21/50  
4/4 [=====] - 0s 33ms/step - loss: 0.2353 -  
val_loss: 0.2196  
Epoch 22/50  
4/4 [=====] - 0s 32ms/step - loss: 0.2085 -  
val_loss: 0.1945  
Epoch 23/50  
4/4 [=====] - 0s 35ms/step - loss: 0.1845 -  
val_loss: 0.1723  
Epoch 24/50  
4/4 [=====] - 0s 36ms/step - loss: 0.1632 -  
val_loss: 0.1527  
Epoch 25/50  
4/4 [=====] - 0s 48ms/step - loss: 0.1447 -  
val_loss: 0.1357  
Epoch 26/50  
4/4 [=====] - 0s 63ms/step - loss: 0.1286 -  
val_loss: 0.1210  
Epoch 27/50  
4/4 [=====] - 0s 49ms/step - loss: 0.1147 -  
val_loss: 0.1083  
Epoch 28/50  
4/4 [=====] - 0s 52ms/step - loss: 0.1028 -  
val_loss: 0.0974  
Epoch 29/50  
4/4 [=====] - 0s 49ms/step - loss: 0.0926 -  
val_loss: 0.0881  
Epoch 30/50  
4/4 [=====] - 0s 52ms/step - loss: 0.0838 -  
val_loss: 0.0800  
Epoch 31/50  
4/4 [=====] - 0s 49ms/step - loss: 0.0763 -
```

```
val_loss: 0.0731
Epoch 32/50
4/4 [=====] - 0s 49ms/step - loss: 0.0698 -
val_loss: 0.0672
Epoch 33/50
4/4 [=====] - 0s 139ms/step - loss: 0.0642 -
val_loss: 0.0620
Epoch 34/50
4/4 [=====] - 0s 72ms/step - loss: 0.0594 -
val_loss: 0.0575
Epoch 35/50
4/4 [=====] - 0s 48ms/step - loss: 0.0552 -
val_loss: 0.0536
Epoch 36/50
4/4 [=====] - 1s 164ms/step - loss: 0.0515 -
val_loss: 0.0502
Epoch 37/50
4/4 [=====] - 0s 37ms/step - loss: 0.0483 -
val_loss: 0.0472
Epoch 38/50
4/4 [=====] - 0s 36ms/step - loss: 0.0454 -
val_loss: 0.0445
Epoch 39/50
4/4 [=====] - 0s 34ms/step - loss: 0.0429 -
val_loss: 0.0421
Epoch 40/50
4/4 [=====] - 0s 33ms/step - loss: 0.0407 -
val_loss: 0.0400
Epoch 41/50
4/4 [=====] - 0s 31ms/step - loss: 0.0387 -
val_loss: 0.0381
Epoch 42/50
4/4 [=====] - 0s 35ms/step - loss: 0.0369 -
val_loss: 0.0364
Epoch 43/50
4/4 [=====] - 0s 32ms/step - loss: 0.0353 -
val_loss: 0.0349
Epoch 44/50
4/4 [=====] - 0s 34ms/step - loss: 0.0339 -
val_loss: 0.0335
Epoch 45/50
4/4 [=====] - 0s 32ms/step - loss: 0.0326 -
val_loss: 0.0323
Epoch 46/50
4/4 [=====] - 0s 40ms/step - loss: 0.0314 -
val_loss: 0.0311
Epoch 47/50
4/4 [=====] - 0s 36ms/step - loss: 0.0303 -
val_loss: 0.0301
```

```
Epoch 48/50
4/4 [=====] - 0s 32ms/step - loss: 0.0293 -
val_loss: 0.0291
Epoch 49/50
4/4 [=====] - 0s 32ms/step - loss: 0.0284 -
val_loss: 0.0282
Epoch 50/50
4/4 [=====] - 0s 33ms/step - loss: 0.0276 -
val_loss: 0.0274
11/11 [=====] - 0s 2ms/step
11/11 [=====] - 0s 3ms/step
```

