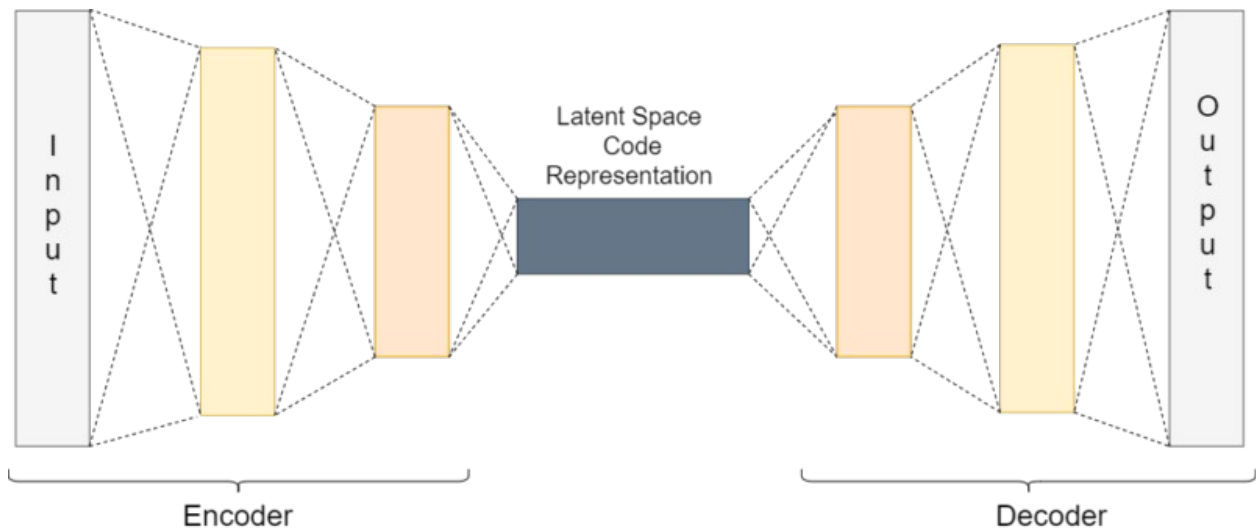


A Brief Introduction to Autoencoders

- Deep learning autoencoders are a type of neural network that can reconstruct specific images from the latent code space.
- The autoencoders obtain the latent code data from a network called the encoder network. Then we give this code as the input to the decoder network which tries to reconstruct the images that the network has been trained on.
- The following image summarizes the above theory in a simple manner.



- The above image summarizes the working of an autoencoder, be it a deep or convolutional autoencoder.
- In one of my previous articles, I have covered the basics of autoencoder in deep learning. You can read the article here ([Autoencoders in Deep Learning](#)).

```
import os
import torch
import torchvision
import torch.nn as nn
import torchvision.transforms as transforms
import torch.optim as optim
import matplotlib.pyplot as plt
import torch.nn.functional as F
import seaborn as sns
from torchvision import datasets
from torch.utils.data import DataLoader
from torchvision.utils import save_image
import os
import pandas as pd
```

```

from PIL import Image
from torchinfo import summary
from torchvision.utils import save_image
from tabulate import tabulate
from tqdm import tqdm
%matplotlib inline

train_image_path="/kaggle/input/10-monkey-species/training/training/"

valid_image_path="/kaggle/input/10-monkey-species/validation/validation/"

```

Visualize Images from Train Data

```

# Get list of class directories
class_dirs = sorted(os.listdir(train_image_path))

# Collect image file paths
image_paths = []
for class_dir in class_dirs:
    class_path = os.path.join(train_image_path, class_dir)
    if os.path.isdir(class_path):
        images = [os.path.join(class_path, img) for img in
os.listdir(class_path)[:5]]
        image_paths.extend(images)

# Select 25 images for plotting
image_paths = image_paths[:25]

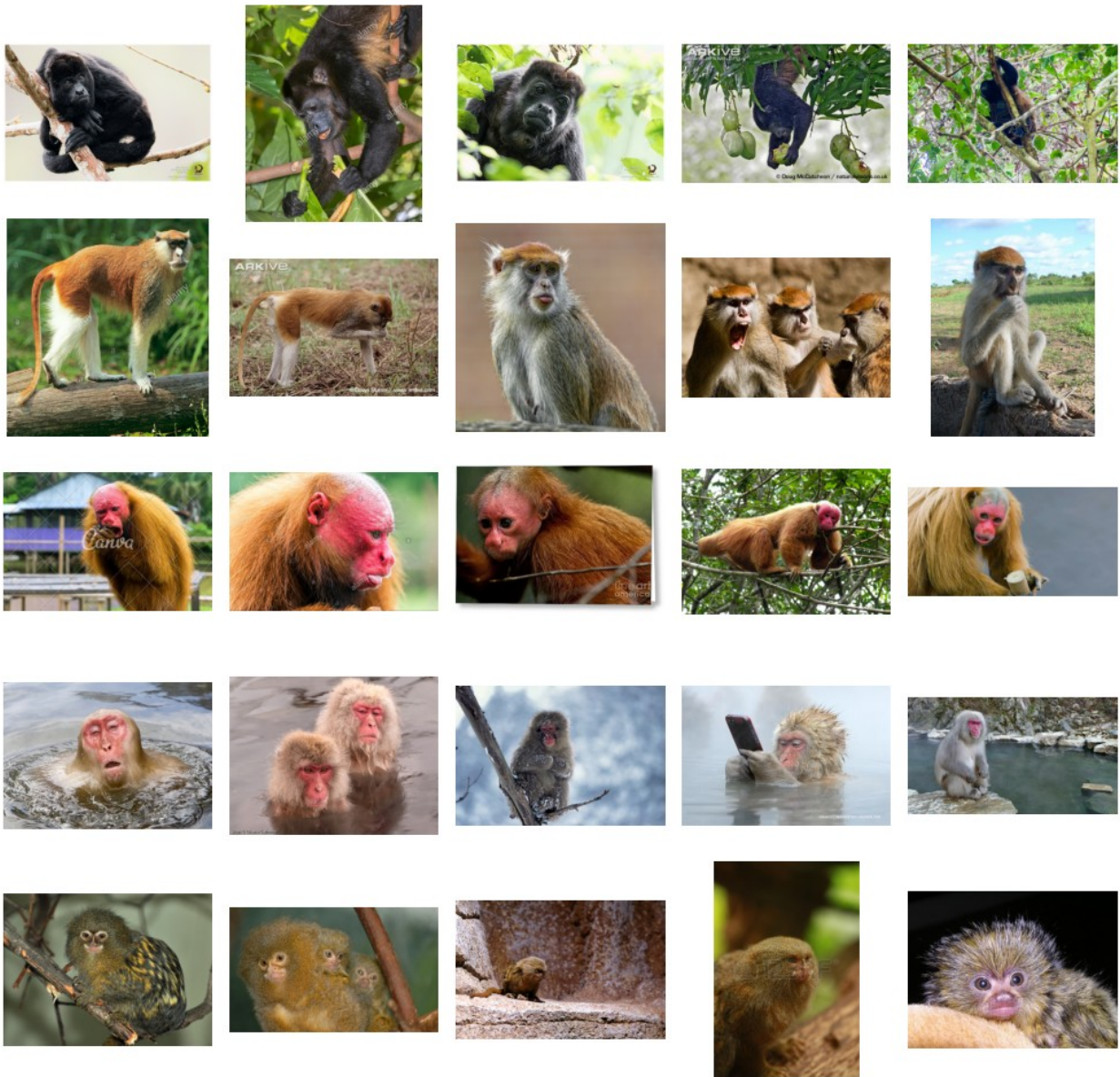
# Plot images in a 5x5 grid
fig, axes = plt.subplots(5, 5, figsize=(10, 10))
fig.suptitle("Sample Images from Training Data", fontsize=16)

for ax, img_path in zip(axes.flat, image_paths):
    img = Image.open(img_path)
    ax.imshow(img)
    ax.axis("off")

plt.tight_layout()
plt.show()

```

Sample Images from Training Data



Visualize Some IMAges from Validation Data

```
# Get list of class directories
class_dirs = sorted(os.listdir(valid_image_path))

# Collect image file paths
image_paths = []
for class_dir in class_dirs:
    class_path = os.path.join(valid_image_path, class_dir)
    if os.path.isdir(class_path):
        images = [os.path.join(class_path, img) for img in
```

```
os.listdir(class_path)[:5]]
    image_paths.extend(images)

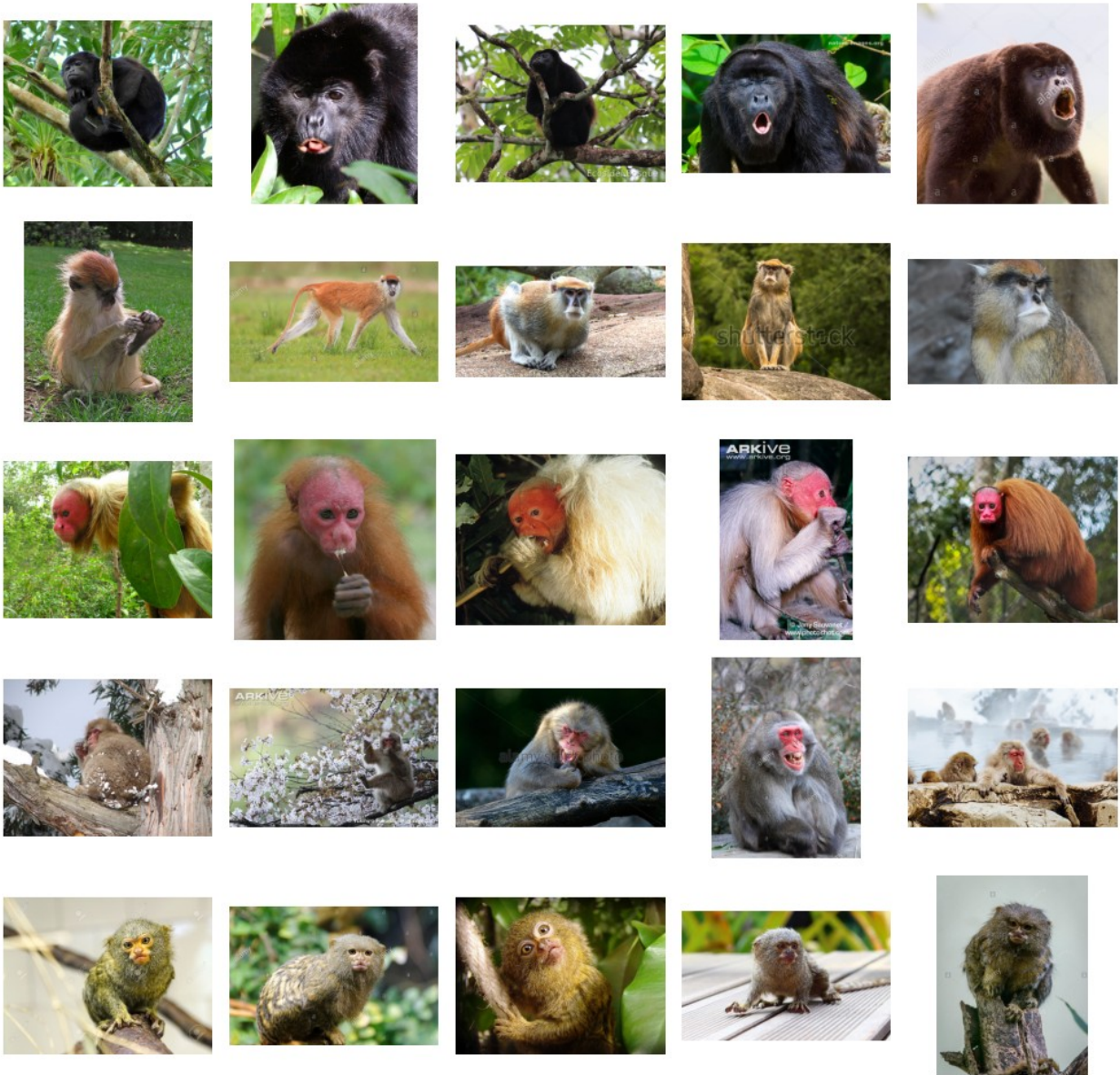
# Select 25 images for plotting
image_paths = image_paths[:25]

# Plot images in a 5x5 grid
fig, axes = plt.subplots(5, 5, figsize=(10, 10))
fig.suptitle("Sample Images from Validation Data", fontsize=16)

for ax, img_path in zip(axes.flat, image_paths):
    img = Image.open(img_path)
    ax.imshow(img)
    ax.axis("off")

plt.tight_layout()
plt.show()
```


Sample Images from Validation Data



```
# Function to count images per class
def count_images_per_class(directory):
    class_counts = {}
    class_dirs = sorted(os.listdir(directory))
    for class_dir in class_dirs:
        class_path = os.path.join(directory, class_dir)
        if os.path.isdir(class_path):
            class_counts[class_dir] = len(os.listdir(class_path))
    return class_counts

# Get image counts
train_counts = count_images_per_class(train_image_path)
```

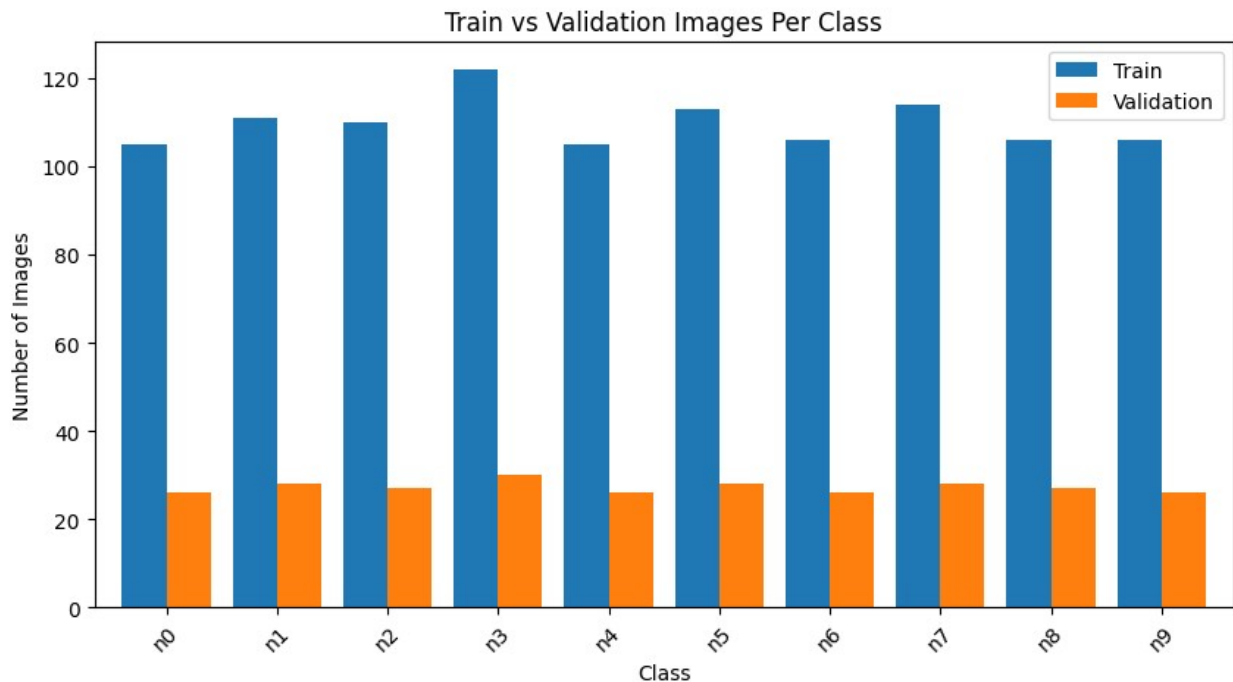
```

valid_counts = count_images_per_class(valid_image_path)

# Create a DataFrame for better visualization
df = pd.DataFrame({'Train': train_counts, 'Validation': valid_counts})
df = df.sort_index()

# Plot the comparison
df.plot(kind='bar', figsize=(10, 5), width=0.8)
plt.title("Train vs Validation Images Per Class")
plt.xlabel("Class")
plt.ylabel("Number of Images")
plt.xticks(rotation=45)
plt.legend(["Train", "Validation"])
plt.show()

```



```

transform = transforms.Compose([
    transforms.Resize((128, 128)), # Resize images to 128x128
    transforms.ToTensor(),          # Convert image to tensor
    transforms.Normalize(mean=[0.5, 0.5, 0.5], std=[0.5, 0.5, 0.5]) #
    Normalize
])

train_dataset = datasets.ImageFolder(root=train_image_path,
transform=transform)
valid_dataset = datasets.ImageFolder(root=valid_image_path,
transform=transform)

```

```

train_loader = DataLoader(train_dataset, batch_size=32, shuffle=True)
valid_loader = DataLoader(valid_dataset, batch_size=32, shuffle=False)

print(f"Train dataset size: {len(train_dataset)} images")
print(f"Validation dataset size: {len(valid_dataset)} images")
print(f"Class labels: {train_dataset.classes}")

Train dataset size: 1097 images
Validation dataset size: 272 images
Class labels: ['n0', 'n1', 'n2', 'n3', 'n4', 'n5', 'n6', 'n7', 'n8', 'n9']

image, label = next(iter(train_loader))
print(f"Image Shape : {image.shape}")
print(f"Label Shape : {label.shape}")

Image Shape : torch.Size([32, 3, 128, 128])
Label Shape : torch.Size([32])

```

Utility Functions

- It is always better to write some utility functions. This would save time and also avoid code repetition. Below are three utility functions that we will need along the way.

```

def get_device():
    if torch.cuda.is_available():
        device = 'cuda:0'
    else:
        device = 'cpu'
    return device

def make_dir():
    image_dir = '/kaggle/working/Monkey_Images'
    if not os.path.exists(image_dir):
        os.makedirs(image_dir)

def save_decoded_image(img, epoch):
    img = img.view(img.size(0), 3, 128, 128) # Correct shape for
model output
    save_image(img, f'./Monkey_Images/linear_ae_image{epoch}.png')

```

- The first function, `get_device()` either returns the GPU device if it is available or the CPU. If you notice, this is a bit different from the one-liner code used in the PyTorch tutorials. This is because some IDEs do not recognize the `torch.device()` method. Therefore, to keep the code compatible for both IDE and python notebooks I just changed the code a bit.

- The second function is `make_dir()` which makes a directory to store the reconstructed images while training. At last, we have `save_decoded_image()` which saves the images that the autoencoder reconstructs.

Define the Autoencoder Network

- In this section, we will define the autoencoder network. Let's define the network first, then we will get to the code explanation.

```
class ConvAutoencoder(nn.Module):

    def __init__(self):
        super(ConvAutoencoder, self).__init__()

        # Encoder
        self.enc1 = nn.Conv2d(3, 64, 3, stride=2, padding=1) #
128x128 -> 64x64
        self.bn1 = nn.BatchNorm2d(64)
        self.enc2 = nn.Conv2d(64, 128, 3, stride=2, padding=1) #
64x64 -> 32x32
        self.bn2 = nn.BatchNorm2d(128)
        self.enc3 = nn.Conv2d(128, 256, 3, stride=2, padding=1) #
32x32 -> 16x16
        self.bn3 = nn.BatchNorm2d(256)
        self.enc4 = nn.Conv2d(256, 512, 3, stride=2, padding=1) #
16x16 -> 8x8
        self.bn4 = nn.BatchNorm2d(512)
        self.enc5 = nn.Conv2d(512, 1024, 3, stride=2, padding=1) #
8x8 -> 4x4
        self.bn5 = nn.BatchNorm2d(1024)

        # Bottleneck
        self.fc1 = nn.Linear(1024 * 4 * 4, 2048) # Flattened
bottleneck
        self.dropout = nn.Dropout(0.3) # Regularization
        self.fc2 = nn.Linear(2048, 1024 * 4 * 4) # Expand back

        # Decoder
        self.dec1 = nn.ConvTranspose2d(1024, 512, 3, stride=2,
padding=1, output_padding=1) # 4x4 -> 8x8
        self.bn6 = nn.BatchNorm2d(512)
        self.dec2 = nn.ConvTranspose2d(512, 256, 3, stride=2,
padding=1, output_padding=1) # 8x8 -> 16x16
        self.bn7 = nn.BatchNorm2d(256)
        self.dec3 = nn.ConvTranspose2d(256, 128, 3, stride=2,
padding=1, output_padding=1) # 16x16 -> 32x32
        self.bn8 = nn.BatchNorm2d(128)
        self.dec4 = nn.ConvTranspose2d(128, 64, 3, stride=2,
padding=1, output_padding=1) # 32x32 -> 64x64
```



```

└─BatchNorm2d: 1-6 [32, 256, 16, 16] 512
└─Conv2d: 1-7 [32, 512, 8, 8]
1,180,160
└─BatchNorm2d: 1-8 [32, 512, 8, 8]
1,024
└─Conv2d: 1-9 [32, 1024, 4, 4]
4,719,616
└─BatchNorm2d: 1-10 [32, 1024, 4, 4]
2,048
└─Linear: 1-11 [32, 2048]
33,556,480
└─Dropout: 1-12 [32, 2048] --
└─Linear: 1-13 [32, 16384]
33,570,816
└─ConvTranspose2d: 1-14 [32, 512, 8, 8]
4,719,104
└─BatchNorm2d: 1-15 [32, 512, 8, 8]
1,024
└─ConvTranspose2d: 1-16 [32, 256, 16, 16]
1,179,904
└─BatchNorm2d: 1-17 [32, 256, 16, 16] 512
└─ConvTranspose2d: 1-18 [32, 128, 32, 32]
295,040
└─BatchNorm2d: 1-19 [32, 128, 32, 32] 256
└─ConvTranspose2d: 1-20 [32, 64, 64, 64]
73,792
└─BatchNorm2d: 1-21 [32, 64, 64, 64] 128
└─ConvTranspose2d: 1-22 [32, 3, 128, 128]
1,731
=====
=====
Total params: 79,673,347
Trainable params: 79,673,347
Non-trainable params: 0
Total mult-adds (G): 51.63
=====
=====
Input size (MB): 6.29
Forward/backward pass size (MB): 529.01
Params size (MB): 318.69
Estimated Total Size (MB): 853.99
=====
=====

```

Define Constants and Prepare the Data

```

NUM_EPOCHS = 100
LEARNING_RATE = 5e-5

```

```
criterion = nn.MSELoss()
optimizer = optim.Adam(model.parameters(), lr=LEARNING_RATE)
```

Train and Test Functions

```
def train(net, trainloader, valloader, num_epochs, optimizer,
criterion, device):
    train_loss = []
    val_loss = []

    # Initialize ReduceLRonPlateau scheduler
    scheduler = torch.optim.lr_scheduler.ReduceLRonPlateau(optimizer,
mode='min', factor=0.5, patience=3, verbose=True, min_lr=1e-6)

    # Early stopping parameters
    best_val_loss = float('inf')
    patience = 5 # Number of epochs to wait before stopping
    patience_counter = 0
    early_stop = False

    for epoch in range(num_epochs):
        if early_stop:
            print("Early stopping triggered.")
            break

        net.train()
        running_train_loss = 0.0
        for data in tqdm(trainloader, desc=f"Training Epoch
{epoch+1}/{num_epochs}", unit="batch"):
            img, _ = data
            img = img.to(device)

            optimizer.zero_grad()
            outputs = net(img)
            loss = criterion(outputs, img)
            loss.backward()
            optimizer.step()
            running_train_loss += loss.item()

        avg_train_loss = running_train_loss / len(trainloader)
        train_loss.append(avg_train_loss)

        net.eval()
        running_val_loss = 0.0
        with torch.no_grad():
            for data in tqdm(valloader, desc=f"Validation Epoch
{epoch+1}/{num_epochs}", unit="batch"):
```

```

        img, _ = data
        img = img.to(device)

        outputs = net(img)
        loss = criterion(outputs, img)
        running_val_loss += loss.item()

    avg_val_loss = running_val_loss / len(valloader)
    val_loss.append(avg_val_loss)

    print(tabulate([[ 'Epoch', epoch+1, 'Train Loss',
f'{avg_train_loss:.3f}', 'Val Loss', f'{avg_val_loss:.3f}' ]],
        headers=[ 'Metric', 'Value', 'Metric', 'Value',
'Metric', 'Value'], tablefmt='grid'))

    # Step the scheduler based on validation loss
    scheduler.step(avg_val_loss)

    # Early stopping logic
    if avg_val_loss < best_val_loss:
        best_val_loss = avg_val_loss
        patience_counter = 0 # Reset counter if validation loss
improves
        # Optionally save the best model
        torch.save(net.state_dict(), 'best_model.pth')
    else:
        patience_counter += 1
        print(f"Validation loss did not improve. Patience counter:
{patience_counter}/{patience}")
        if patience_counter >= patience:
            early_stop = True

    if (epoch + 1) % 5 == 0:
        save_decoded_image(outputs.cpu().data, epoch + 1)

    return train_loss, val_loss

def test_image_reconstruction(net, testloader, device):
    for batch in tqdm(testloader, desc="Testing", unit="batch"):
        img, _ = batch
        img = img.to(device)

        outputs = net(img)
        outputs = outputs.view(outputs.size(0), 3, 128,
128).cpu().data # Fixed to match model output

        save_image(outputs, 'monkey_reconstruction.png')
        break

# Assuming model, train_loader, valid_loader, optimizer, criterion,

```

```
and NUM_EPOCHS are defined
device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
model.to(device)
```

```
num_epochs = NUM_EPOCHS # Assuming this is defined elsewhere
```

```
make_dir()
```

```
print("Training the model...")
```

```
train_loss, val_loss = train(model, train_loader, valid_loader,
num_epochs, optimizer, criterion, device)
```

```
/usr/local/lib/python3.10/dist-packages/torch/optim/
lr_scheduler.py:62: UserWarning: The verbose parameter is deprecated.
Please use get_last_lr() to access the learning rate.
  warnings.warn(
```

```
Training the model...
```

```
Training Epoch 1/100: 100%|██████████| 35/35 [00:24<00:00,
1.45batch/s]
```

```
Validation Epoch 1/100: 100%|██████████| 9/9 [00:06<00:00,
1.43batch/s]
```

+	-----+	-----+	-----+	-----+	-----+	-----+						
	Metric		Value		Metric		Value		Metric		Value	
+	=====+	=====+	=====+	=====+	=====+	=====+						
	Epoch		1		Train Loss		0.687		Val Loss		0.661	
+	-----+	-----+	-----+	-----+	-----+	-----+						

```
Training Epoch 2/100: 100%|██████████| 35/35 [00:23<00:00,
1.47batch/s]
```

```
Validation Epoch 2/100: 100%|██████████| 9/9 [00:06<00:00,
1.45batch/s]
```

+	-----+	-----+	-----+	-----+	-----+	-----+						
	Metric		Value		Metric		Value		Metric		Value	
+	=====+	=====+	=====+	=====+	=====+	=====+						
	Epoch		2		Train Loss		0.598		Val Loss		0.58	
+	-----+	-----+	-----+	-----+	-----+	-----+						

```
Training Epoch 3/100: 100%|██████████| 35/35 [00:23<00:00,
1.46batch/s]
```

```
Validation Epoch 3/100: 100%|██████████| 9/9 [00:06<00:00,
1.43batch/s]
```

+	-----+	-----+	-----+	-----+	-----+	-----+						
	Metric		Value		Metric		Value		Metric		Value	
+	=====+	=====+	=====+	=====+	=====+	=====+						
	Epoch		3		Train Loss		0.536		Val Loss		0.514	
+	-----+	-----+	-----+	-----+	-----+	-----+						

Training Epoch 4/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 4/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	4	Train Loss	0.504	Val Loss	0.508

Training Epoch 5/100: 100%|██████████| 35/35 [00:23<00:00, 1.46batch/s]
Validation Epoch 5/100: 100%|██████████| 9/9 [00:06<00:00, 1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	5	Train Loss	0.475	Val Loss	0.473

Training Epoch 6/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]
Validation Epoch 6/100: 100%|██████████| 9/9 [00:06<00:00, 1.47batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	6	Train Loss	0.458	Val Loss	0.469

Training Epoch 7/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 7/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	7	Train Loss	0.442	Val Loss	0.451

Training Epoch 8/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 8/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
--------	-------	--------	-------	--------	-------

Epoch	8	Train Loss	0.425	Val Loss	0.429
-------	---	------------	-------	----------	-------

Training Epoch 9/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 9/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	9	Train Loss	0.412	Val Loss	0.409

Training Epoch 10/100: 100%|██████████| 35/35 [00:23<00:00, 1.49batch/s]
Validation Epoch 10/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	10	Train Loss	0.398	Val Loss	0.399

Training Epoch 11/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 11/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	11	Train Loss	0.386	Val Loss	0.381

Training Epoch 12/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 12/100: 100%|██████████| 9/9 [00:06<00:00, 1.41batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	12	Train Loss	0.378	Val Loss	0.375

Training Epoch 13/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]

Validation Epoch 13/100: 100%|██████████| 9/9 [00:06<00:00, 1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	13	Train Loss	0.369	Val Loss	0.358

Training Epoch 14/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]

Validation Epoch 14/100: 100%|██████████| 9/9 [00:06<00:00, 1.43batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	14	Train Loss	0.36	Val Loss	0.356

Training Epoch 15/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]

Validation Epoch 15/100: 100%|██████████| 9/9 [00:06<00:00, 1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	15	Train Loss	0.351	Val Loss	0.351

Training Epoch 16/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]

Validation Epoch 16/100: 100%|██████████| 9/9 [00:06<00:00, 1.42batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	16	Train Loss	0.346	Val Loss	0.345

Training Epoch 17/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]

Validation Epoch 17/100: 100%|██████████| 9/9 [00:06<00:00, 1.43batch/s]

Metric	Value	Metric	Value	Metric	Value
--------	-------	--------	-------	--------	-------

Epoch		17	Train Loss		0.34	Val Loss		0.34	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 18/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]
Validation Epoch 18/100: 100%|██████████| 9/9 [00:06<00:00, 1.40batch/s]

Metric		Value	Metric		Value	Metric		Value	
+=====+=====+=====+=====+=====+=====+=====+=====+									
Epoch		18	Train Loss		0.332	Val Loss		0.328	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 19/100: 100%|██████████| 35/35 [00:24<00:00, 1.43batch/s]
Validation Epoch 19/100: 100%|██████████| 9/9 [00:06<00:00, 1.36batch/s]

Metric		Value	Metric		Value	Metric		Value	
+=====+=====+=====+=====+=====+=====+=====+=====+									
Epoch		19	Train Loss		0.328	Val Loss		0.326	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 20/100: 100%|██████████| 35/35 [00:23<00:00, 1.46batch/s]
Validation Epoch 20/100: 100%|██████████| 9/9 [00:06<00:00, 1.42batch/s]

Metric		Value	Metric		Value	Metric		Value	
+=====+=====+=====+=====+=====+=====+=====+=====+									
Epoch		20	Train Loss		0.322	Val Loss		0.322	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 21/100: 100%|██████████| 35/35 [00:24<00:00, 1.45batch/s]
Validation Epoch 21/100: 100%|██████████| 9/9 [00:06<00:00, 1.41batch/s]

Metric		Value	Metric		Value	Metric		Value	
+=====+=====+=====+=====+=====+=====+=====+=====+									
Epoch		21	Train Loss		0.318	Val Loss		0.318	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 22/100: 100%|██████████| 35/35 [00:24<00:00, 1.46batch/s]
Validation Epoch 22/100: 100%|██████████| 9/9 [00:06<00:00, 1.40batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	22	Train Loss	0.312	Val Loss	0.307

Training Epoch 23/100: 100%|██████████| 35/35 [00:23<00:00, 1.46batch/s]
Validation Epoch 23/100: 100%|██████████| 9/9 [00:06<00:00, 1.42batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	23	Train Loss	0.308	Val Loss	0.307

Validation loss did not improve. Patience counter: 1/5

Training Epoch 24/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 24/100: 100%|██████████| 9/9 [00:06<00:00, 1.43batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	24	Train Loss	0.304	Val Loss	0.304

Training Epoch 25/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]
Validation Epoch 25/100: 100%|██████████| 9/9 [00:06<00:00, 1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	25	Train Loss	0.3	Val Loss	0.303

Training Epoch 26/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]
Validation Epoch 26/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	26	Train Loss	0.299	Val Loss	0.301

Training Epoch 27/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 27/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	27	Train Loss	0.295	Val Loss	0.298

Training Epoch 28/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 28/100: 100%|██████████| 9/9 [00:06<00:00, 1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	28	Train Loss	0.292	Val Loss	0.293

Training Epoch 29/100: 100%|██████████| 35/35 [00:23<00:00, 1.46batch/s]
Validation Epoch 29/100: 100%|██████████| 9/9 [00:06<00:00, 1.43batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	29	Train Loss	0.288	Val Loss	0.292

Training Epoch 30/100: 100%|██████████| 35/35 [00:23<00:00, 1.46batch/s]
Validation Epoch 30/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	30	Train Loss	0.285	Val Loss	0.289

Training Epoch 31/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 31/100: 100%|██████████| 9/9 [00:06<00:00, 1.47batch/s]

Metric	Value	Metric	Value	Metric	Value
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Epoch	31	Train Loss	0.284	Val Loss	0.283
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Training Epoch 32/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]

Validation Epoch 32/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	32	Train Loss	0.283	Val Loss	0.287

Validation loss did not improve. Patience counter: 1/5

Training Epoch 33/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]

Validation Epoch 33/100: 100%|██████████| 9/9 [00:06<00:00, 1.47batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	33	Train Loss	0.28	Val Loss	0.286

Validation loss did not improve. Patience counter: 2/5

Training Epoch 34/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]

Validation Epoch 34/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	34	Train Loss	0.277	Val Loss	0.28

Training Epoch 35/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]

Validation Epoch 35/100: 100%|██████████| 9/9 [00:06<00:00, 1.47batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	35	Train Loss	0.275	Val Loss	0.28

Validation loss did not improve. Patience counter: 1/5

Training Epoch 36/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]
Validation Epoch 36/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	36	Train Loss	0.272	Val Loss	0.276

Training Epoch 37/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]
Validation Epoch 37/100: 100%|██████████| 9/9 [00:06<00:00, 1.47batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	37	Train Loss	0.272	Val Loss	0.277

Validation loss did not improve. Patience counter: 1/5

Training Epoch 38/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 38/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	38	Train Loss	0.269	Val Loss	0.273

Training Epoch 39/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 39/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	39	Train Loss	0.269	Val Loss	0.275

Validation loss did not improve. Patience counter: 1/5

Training Epoch 40/100: 100%|██████████| 35/35 [00:23<00:00, 1.46batch/s]
Validation Epoch 40/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	40	Train Loss	0.267	Val Loss	0.275

Validation loss did not improve. Patience counter: 2/5

Training Epoch 41/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]

Validation Epoch 41/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	41	Train Loss	0.266	Val Loss	0.271

Training Epoch 42/100: 100%|██████████| 35/35 [00:23<00:00, 1.46batch/s]

Validation Epoch 42/100: 100%|██████████| 9/9 [00:06<00:00, 1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	42	Train Loss	0.264	Val Loss	0.273

Validation loss did not improve. Patience counter: 1/5

Training Epoch 43/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]

Validation Epoch 43/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	43	Train Loss	0.265	Val Loss	0.272

Validation loss did not improve. Patience counter: 2/5

Training Epoch 44/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]

Validation Epoch 44/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
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Epoch		44	Train Loss		0.263	Val Loss		0.267	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 45/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 45/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric		Value	Metric		Value	Metric		Value	
+=====+=====+=====+=====+=====+=====+=====+=====+									
Epoch		45	Train Loss		0.258	Val Loss		0.266	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 46/100: 100%|██████████| 35/35 [00:24<00:00, 1.44batch/s]
Validation Epoch 46/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric		Value	Metric		Value	Metric		Value	
+=====+=====+=====+=====+=====+=====+=====+=====+									
Epoch		46	Train Loss		0.257	Val Loss		0.264	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 47/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 47/100: 100%|██████████| 9/9 [00:06<00:00, 1.36batch/s]

Metric		Value	Metric		Value	Metric		Value	
+=====+=====+=====+=====+=====+=====+=====+=====+									
Epoch		47	Train Loss		0.258	Val Loss		0.265	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Validation loss did not improve. Patience counter: 1/5

Training Epoch 48/100: 100%|██████████| 35/35 [00:24<00:00, 1.42batch/s]
Validation Epoch 48/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric		Value	Metric		Value	Metric		Value	
+=====+=====+=====+=====+=====+=====+=====+=====+									
Epoch		48	Train Loss		0.257	Val Loss		0.263	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 49/100: 100%|██████████| 35/35 [00:24<00:00, 1.45batch/s]

Validation Epoch 49/100: 100%|██████████| 9/9 [00:06<00:00, 1.43batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	49	Train Loss	0.256	Val Loss	0.263

Training Epoch 50/100: 100%|██████████| 35/35 [00:24<00:00, 1.44batch/s]

Validation Epoch 50/100: 100%|██████████| 9/9 [00:06<00:00, 1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	50	Train Loss	0.254	Val Loss	0.262

Training Epoch 51/100: 100%|██████████| 35/35 [00:24<00:00, 1.45batch/s]

Validation Epoch 51/100: 100%|██████████| 9/9 [00:06<00:00, 1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	51	Train Loss	0.253	Val Loss	0.262

Training Epoch 52/100: 100%|██████████| 35/35 [00:24<00:00, 1.44batch/s]

Validation Epoch 52/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	52	Train Loss	0.254	Val Loss	0.261

Training Epoch 53/100: 100%|██████████| 35/35 [00:24<00:00, 1.46batch/s]

Validation Epoch 53/100: 100%|██████████| 9/9 [00:06<00:00, 1.41batch/s]

Metric	Value	Metric	Value	Metric	Value
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Epoch		53	Train Loss		0.252	Val Loss		0.26	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 54/100: 100%|██████████| 35/35 [00:24<00:00, 1.46batch/s]
Validation Epoch 54/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric		Value	Metric		Value	Metric		Value	
+=====+=====+=====+=====+=====+=====+=====+=====+									
Epoch		54	Train Loss		0.253	Val Loss		0.26	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 55/100: 100%|██████████| 35/35 [00:24<00:00, 1.45batch/s]
Validation Epoch 55/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric		Value	Metric		Value	Metric		Value	
+=====+=====+=====+=====+=====+=====+=====+=====+									
Epoch		55	Train Loss		0.25	Val Loss		0.258	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 56/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 56/100: 100%|██████████| 9/9 [00:06<00:00, 1.47batch/s]

Metric		Value	Metric		Value	Metric		Value	
+=====+=====+=====+=====+=====+=====+=====+=====+									
Epoch		56	Train Loss		0.248	Val Loss		0.258	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 57/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 57/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric		Value	Metric		Value	Metric		Value	
+=====+=====+=====+=====+=====+=====+=====+=====+									
Epoch		57	Train Loss		0.248	Val Loss		0.258	
+-----+-----+-----+-----+-----+-----+-----+-----+									

Training Epoch 58/100: 100%|██████████| 35/35 [00:24<00:00, 1.46batch/s]
Validation Epoch 58/100: 100%|██████████| 9/9 [00:06<00:00, 1.47batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	58	Train Loss	0.247	Val Loss	0.255

Training Epoch 59/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 59/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	59	Train Loss	0.246	Val Loss	0.254

Training Epoch 60/100: 100%|██████████| 35/35 [00:24<00:00, 1.46batch/s]
Validation Epoch 60/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	60	Train Loss	0.244	Val Loss	0.254

Validation loss did not improve. Patience counter: 1/5

Training Epoch 61/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 61/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	61	Train Loss	0.245	Val Loss	0.255

Validation loss did not improve. Patience counter: 2/5

Training Epoch 62/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 62/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	62	Train Loss	0.243	Val Loss	0.253

Training Epoch 63/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 63/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	63	Train Loss	0.244	Val Loss	0.254

Validation loss did not improve. Patience counter: 1/5

Training Epoch 64/100: 100%|██████████| 35/35 [00:23<00:00, 1.49batch/s]
Validation Epoch 64/100: 100%|██████████| 9/9 [00:06<00:00, 1.40batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	64	Train Loss	0.243	Val Loss	0.252

Training Epoch 65/100: 100%|██████████| 35/35 [00:23<00:00, 1.46batch/s]
Validation Epoch 65/100: 100%|██████████| 9/9 [00:06<00:00, 1.38batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	65	Train Loss	0.243	Val Loss	0.251

Training Epoch 66/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 66/100: 100%|██████████| 9/9 [00:06<00:00, 1.40batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	66	Train Loss	0.241	Val Loss	0.252

Validation loss did not improve. Patience counter: 1/5

Training Epoch 67/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 67/100: 100%|██████████| 9/9 [00:06<00:00, 1.39batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	67	Train Loss	0.241	Val Loss	0.254

Validation loss did not improve. Patience counter: 2/5

Training Epoch 68/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]

Validation Epoch 68/100: 100%|██████████| 9/9 [00:06<00:00, 1.42batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	68	Train Loss	0.241	Val Loss	0.251

Training Epoch 69/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]

Validation Epoch 69/100: 100%|██████████| 9/9 [00:06<00:00, 1.41batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	69	Train Loss	0.24	Val Loss	0.25

Training Epoch 70/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]

Validation Epoch 70/100: 100%|██████████| 9/9 [00:06<00:00, 1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	70	Train Loss	0.238	Val Loss	0.25

Training Epoch 71/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]

Validation Epoch 71/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	71	Train Loss	0.24	Val Loss	0.25

Validation loss did not improve. Patience counter: 1/5

Training Epoch 72/100: 100%|██████████| 35/35 [00:23<00:00, 1.49batch/s]
Validation Epoch 72/100: 100%|██████████| 9/9 [00:06<00:00, 1.47batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	72	Train Loss	0.236	Val Loss	0.248

Training Epoch 73/100: 100%|██████████| 35/35 [00:23<00:00, 1.46batch/s]
Validation Epoch 73/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	73	Train Loss	0.236	Val Loss	0.248

Training Epoch 74/100: 100%|██████████| 35/35 [00:23<00:00, 1.46batch/s]
Validation Epoch 74/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	74	Train Loss	0.235	Val Loss	0.247

Training Epoch 75/100: 100%|██████████| 35/35 [00:24<00:00, 1.44batch/s]
Validation Epoch 75/100: 100%|██████████| 9/9 [00:06<00:00, 1.43batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	75	Train Loss	0.236	Val Loss	0.248

Validation loss did not improve. Patience counter: 1/5

Training Epoch 76/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 76/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	76	Train Loss	0.235	Val Loss	0.248

Validation loss did not improve. Patience counter: 2/5

Training Epoch 77/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]

Validation Epoch 77/100: 100%|██████████| 9/9 [00:06<00:00, 1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	77	Train Loss	0.233	Val Loss	0.248

Validation loss did not improve. Patience counter: 3/5

Training Epoch 78/100: 100%|██████████| 35/35 [00:23<00:00, 1.49batch/s]

Validation Epoch 78/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	78	Train Loss	0.234	Val Loss	0.246

Training Epoch 79/100: 100%|██████████| 35/35 [00:24<00:00, 1.46batch/s]

Validation Epoch 79/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	79	Train Loss	0.233	Val Loss	0.245

Training Epoch 80/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]

Validation Epoch 80/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	80	Train Loss	0.234	Val Loss	0.246

+-----+-----+-----+-----+-----+
Validation loss did not improve. Patience counter: 1/5

Training Epoch 81/100: 100%|██████████| 35/35 [00:23<00:00,
1.47batch/s]

Validation Epoch 81/100: 100%|██████████| 9/9 [00:06<00:00,
1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	81	Train Loss	0.232	Val Loss	0.246

+-----+-----+-----+-----+-----+
Validation loss did not improve. Patience counter: 2/5

Training Epoch 82/100: 100%|██████████| 35/35 [00:23<00:00,
1.48batch/s]

Validation Epoch 82/100: 100%|██████████| 9/9 [00:06<00:00,
1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	82	Train Loss	0.232	Val Loss	0.247

+-----+-----+-----+-----+-----+
Validation loss did not improve. Patience counter: 3/5

Training Epoch 83/100: 100%|██████████| 35/35 [00:23<00:00,
1.47batch/s]

Validation Epoch 83/100: 100%|██████████| 9/9 [00:06<00:00,
1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	83	Train Loss	0.232	Val Loss	0.246

+-----+-----+-----+-----+-----+
Validation loss did not improve. Patience counter: 4/5

Training Epoch 84/100: 100%|██████████| 35/35 [00:23<00:00,
1.47batch/s]

Validation Epoch 84/100: 100%|██████████| 9/9 [00:06<00:00,
1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	84	Train Loss	0.23	Val Loss	0.245

Training Epoch 85/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 85/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	85	Train Loss	0.231	Val Loss	0.246

Validation loss did not improve. Patience counter: 1/5

Training Epoch 86/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 86/100: 100%|██████████| 9/9 [00:06<00:00, 1.47batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	86	Train Loss	0.23	Val Loss	0.245

Validation loss did not improve. Patience counter: 2/5

Training Epoch 87/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]
Validation Epoch 87/100: 100%|██████████| 9/9 [00:06<00:00, 1.48batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	87	Train Loss	0.23	Val Loss	0.245

Validation loss did not improve. Patience counter: 3/5

Training Epoch 88/100: 100%|██████████| 35/35 [00:23<00:00, 1.49batch/s]
Validation Epoch 88/100: 100%|██████████| 9/9 [00:06<00:00, 1.48batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	88	Train Loss	0.23	Val Loss	0.245

Validation loss did not improve. Patience counter: 4/5

Training Epoch 89/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]

Validation Epoch 89/100: 100%|██████████| 9/9 [00:06<00:00, 1.45batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	89	Train Loss	0.229	Val Loss	0.244

Training Epoch 90/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]

Validation Epoch 90/100: 100%|██████████| 9/9 [00:06<00:00, 1.46batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	90	Train Loss	0.23	Val Loss	0.245

Validation loss did not improve. Patience counter: 1/5

Training Epoch 91/100: 100%|██████████| 35/35 [00:23<00:00, 1.48batch/s]

Validation Epoch 91/100: 100%|██████████| 9/9 [00:06<00:00, 1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	91	Train Loss	0.23	Val Loss	0.245

Validation loss did not improve. Patience counter: 2/5

Training Epoch 92/100: 100%|██████████| 35/35 [00:23<00:00, 1.49batch/s]

Validation Epoch 92/100: 100%|██████████| 9/9 [00:06<00:00, 1.44batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	92	Train Loss	0.23	Val Loss	0.245

Validation loss did not improve. Patience counter: 3/5

Training Epoch 93/100: 100%|██████████| 35/35 [00:23<00:00, 1.47batch/s]

Validation Epoch 93/100: 100%|██████████| 9/9 [00:06<00:00, 1.38batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	93	Train Loss	0.23	Val Loss	0.245

Validation loss did not improve. Patience counter: 4/5

Training Epoch 94/100: 100%|██████████| 35/35 [00:23<00:00, 1.49batch/s]

Validation Epoch 94/100: 100%|██████████| 9/9 [00:06<00:00, 1.43batch/s]

Metric	Value	Metric	Value	Metric	Value
Epoch	94	Train Loss	0.229	Val Loss	0.244

Validation loss did not improve. Patience counter: 5/5

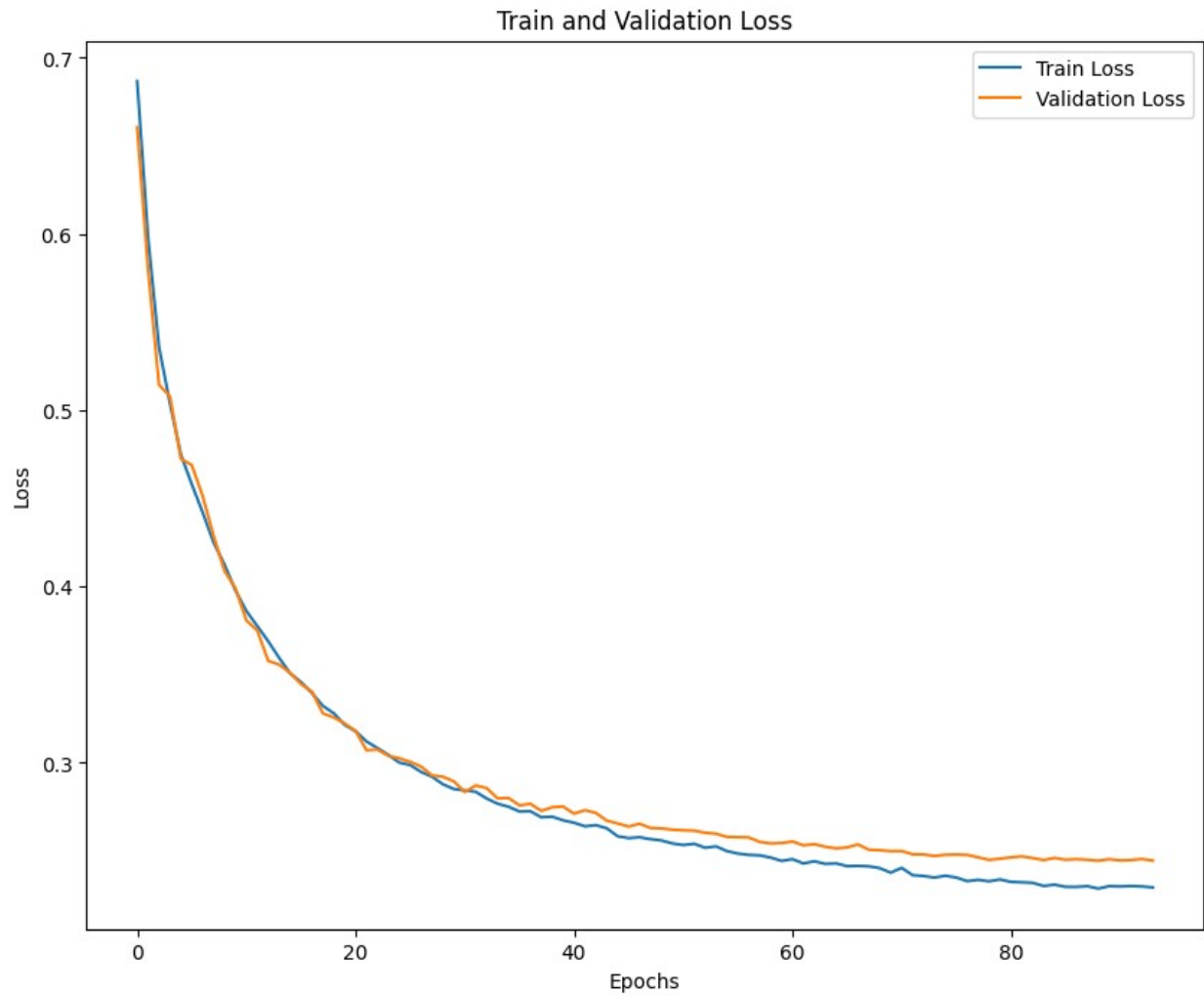
Early stopping triggered.

```
plt.figure(figsize=(10,8))
plt.plot(train_loss, label='Train Loss')
plt.plot(val_loss, label='Validation Loss')
plt.title('Train and Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
plt.savefig('train_val_loss.png')

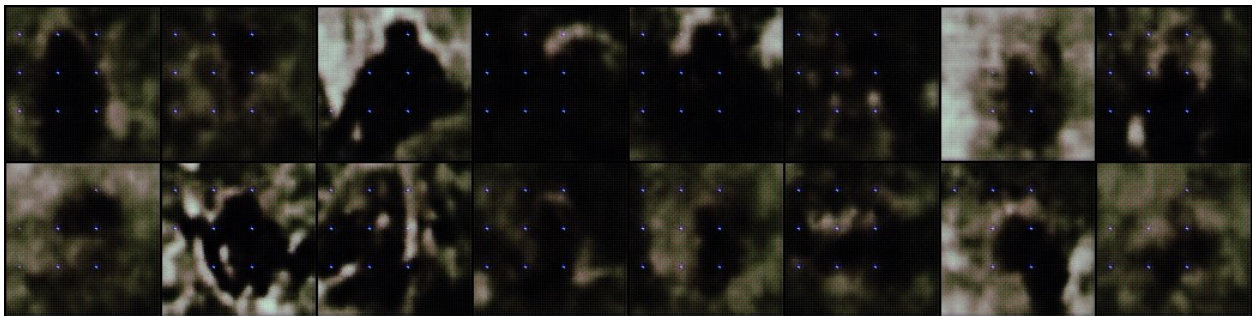
print("Testing the model...")
test_image_reconstruction(model, valid_loader, device)
```

Testing the model...

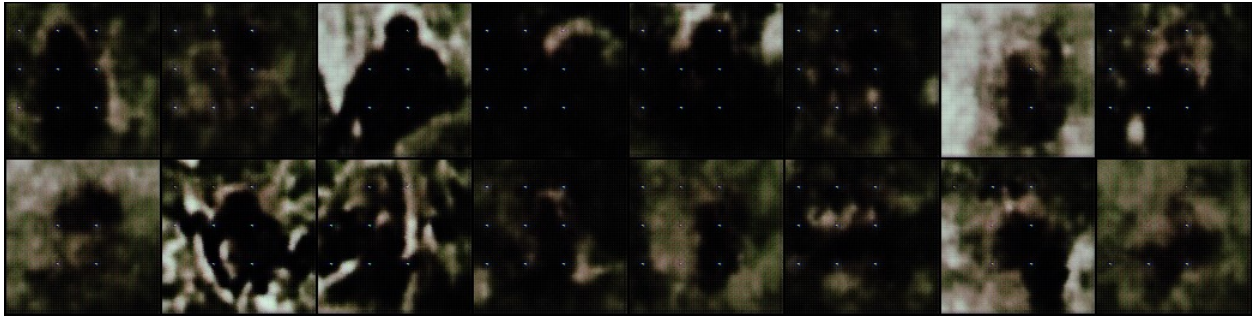
Testing: 0%|██████████| 0/9 [00:01<?, ?batch/s]



```
from IPython.display import Image  
Image('/kaggle/working/Monkey_Images/linear_ae_image50.png')
```



```
Image('/kaggle/working/Monkey_Images/linear_ae_image60.png')
```



```
Image('/kaggle/working/Monkey_Images/linear_ae_image90.png')
```



Reconstructed Images

```
Image("/kaggle/working/monkey_reconstruction.png")
```



Original Images

```
import numpy as np

def imshow(img):
    npimg = img.numpy()
    plt.figure(figsize=(20, 12))
    plt.imshow(np.transpose(npimg, (1, 2, 0)))
    plt.show()

# Get some random training images
dataiter = iter(train_loader)
images, labels = next(dataiter) # Use next() instead of
dataiter.next()

# Show images
imshow(torchvision.utils.make_grid(images))
```

